

# Grid Integration of Variable Renewable Energy

Positive and negative balancing power needs for system operation and planning

Albrecht Tiedemann

1. Disturbed balance between generation and consumption
2. Types of reserves and balancing power calculation
3. Balancing power calculation with probabilistic approach and common distribution function of imbalances
  - Power plant outage distribution
  - Load forecast error distribution
4. Balancing power calculation with probabilistic approach - exercises with excel tool
5. Dynamic balancing power calculation example

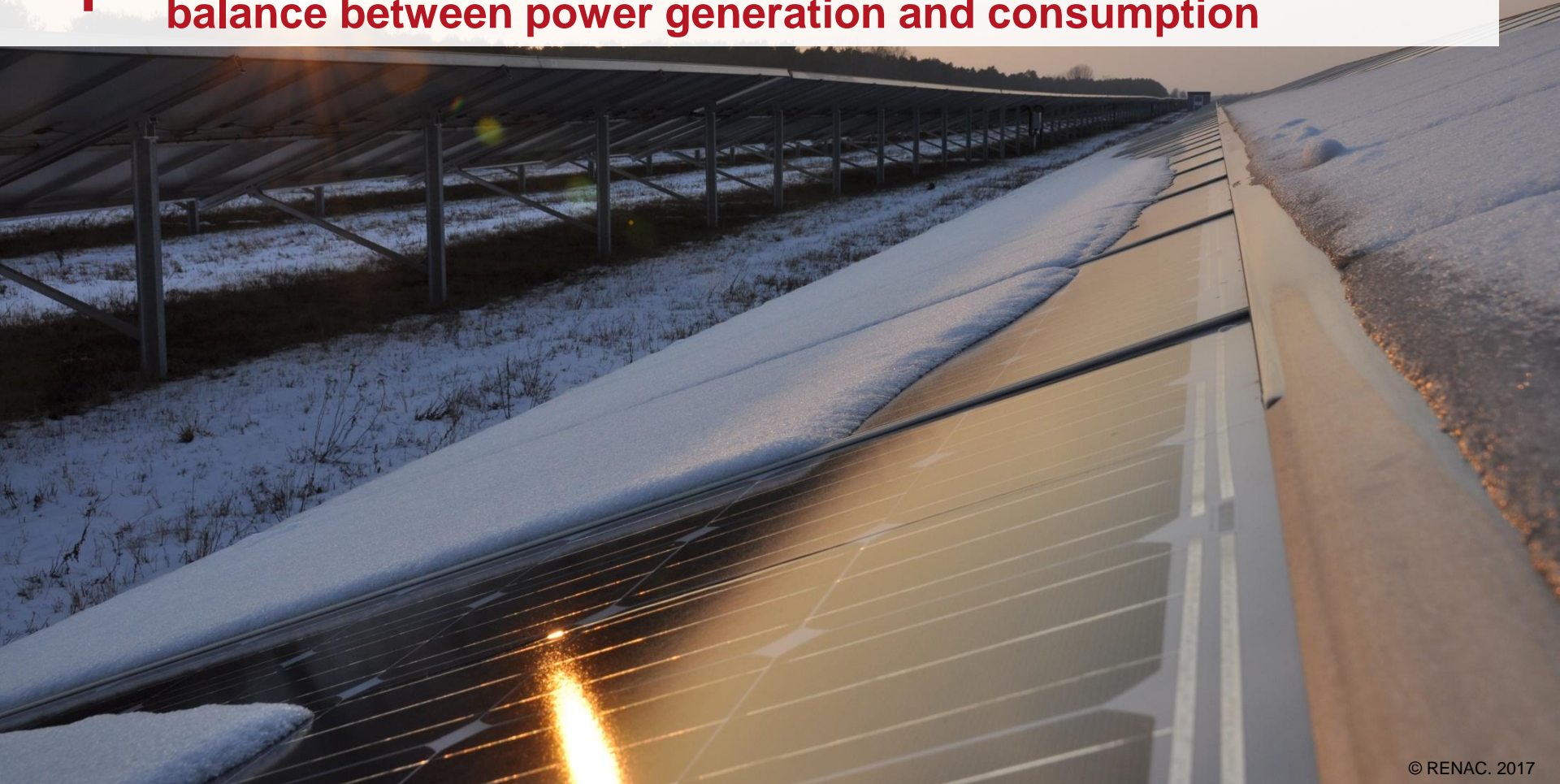
AGENDA

# Learning objectives

- A participant who has met the objectives of the course will be able to
  - name reasons for power imbalances and balancing power needs
  - define types of reserves and balancing power requirements
  - combine different distribution functions of imbalances to dimension balancing power needs
  - use a probabilistic balancing power calculation tool to calculate the amount of positive and negative balancing power needed to ensure a certain system reliability

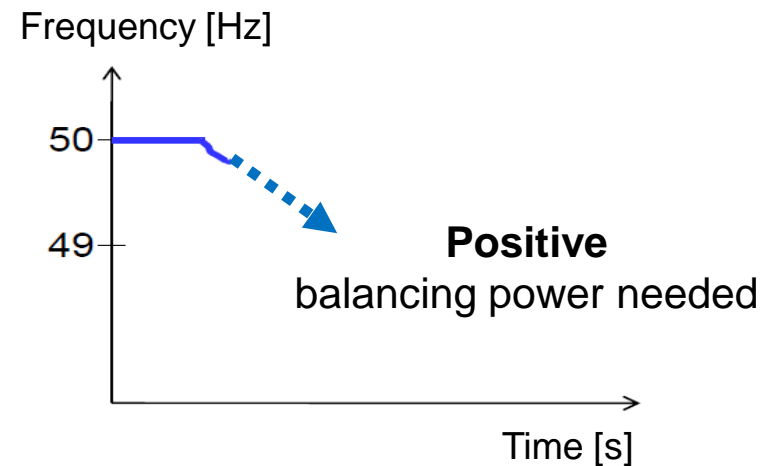
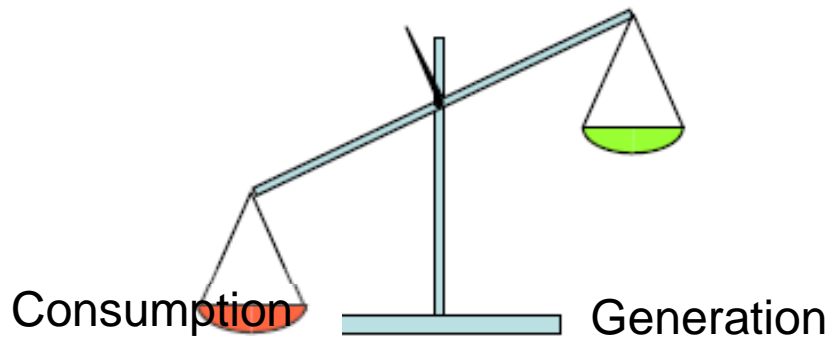
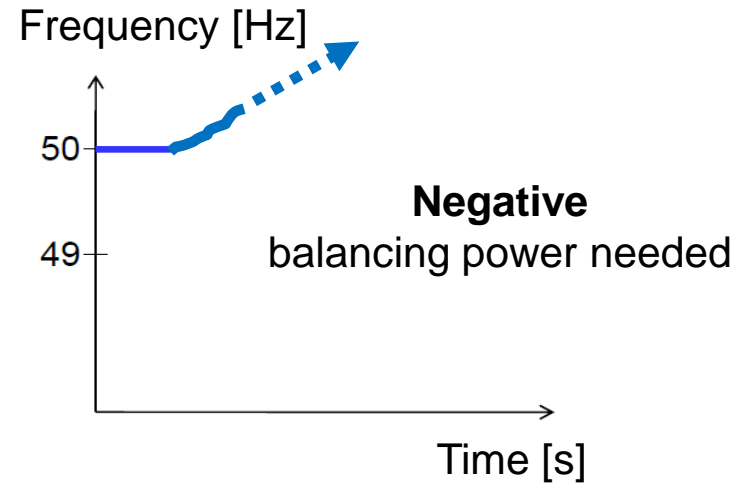
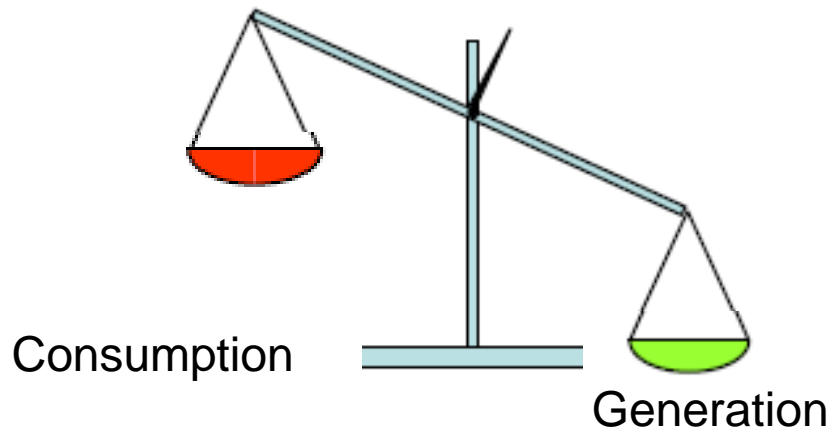
Learning objectives

# 1 Positive and negative reserves to manage disturbed balance between power generation and consumption



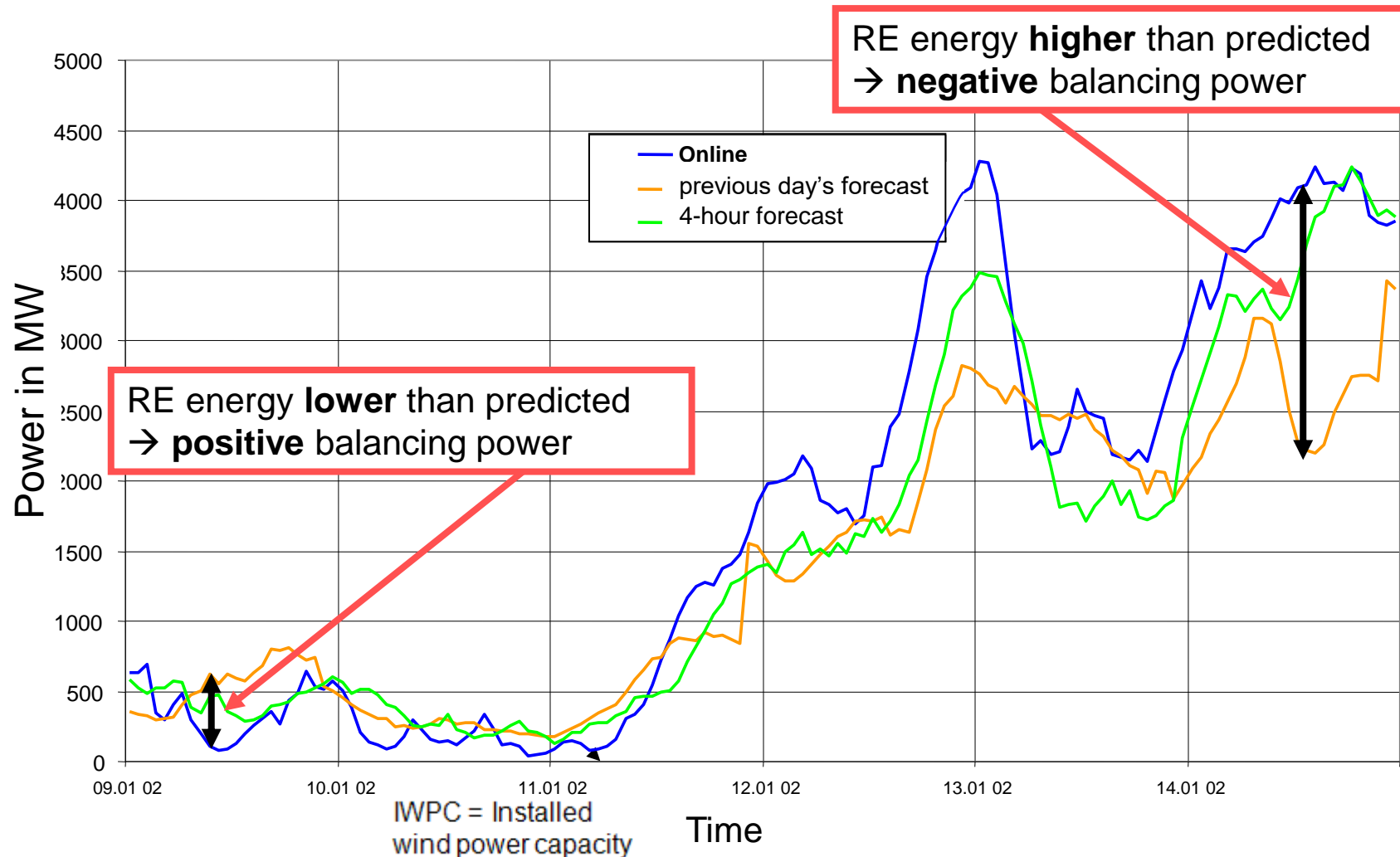


# Disturbed balance between generation and consumption and positive/negative balancing power needs



- Power generation:
  - Unexpected blackout of power stations
  - Deviation of power stations from schedules
  - Short term forecast errors of wind power and solar power
  - Gradient of wind power and solar power feed in
- Load forecast error, load noise and gradient of load
- Outage of transmission and distribution elements

# Day ahead and 4 hour ahead wind forecast and online data, positive / negative balancing power need



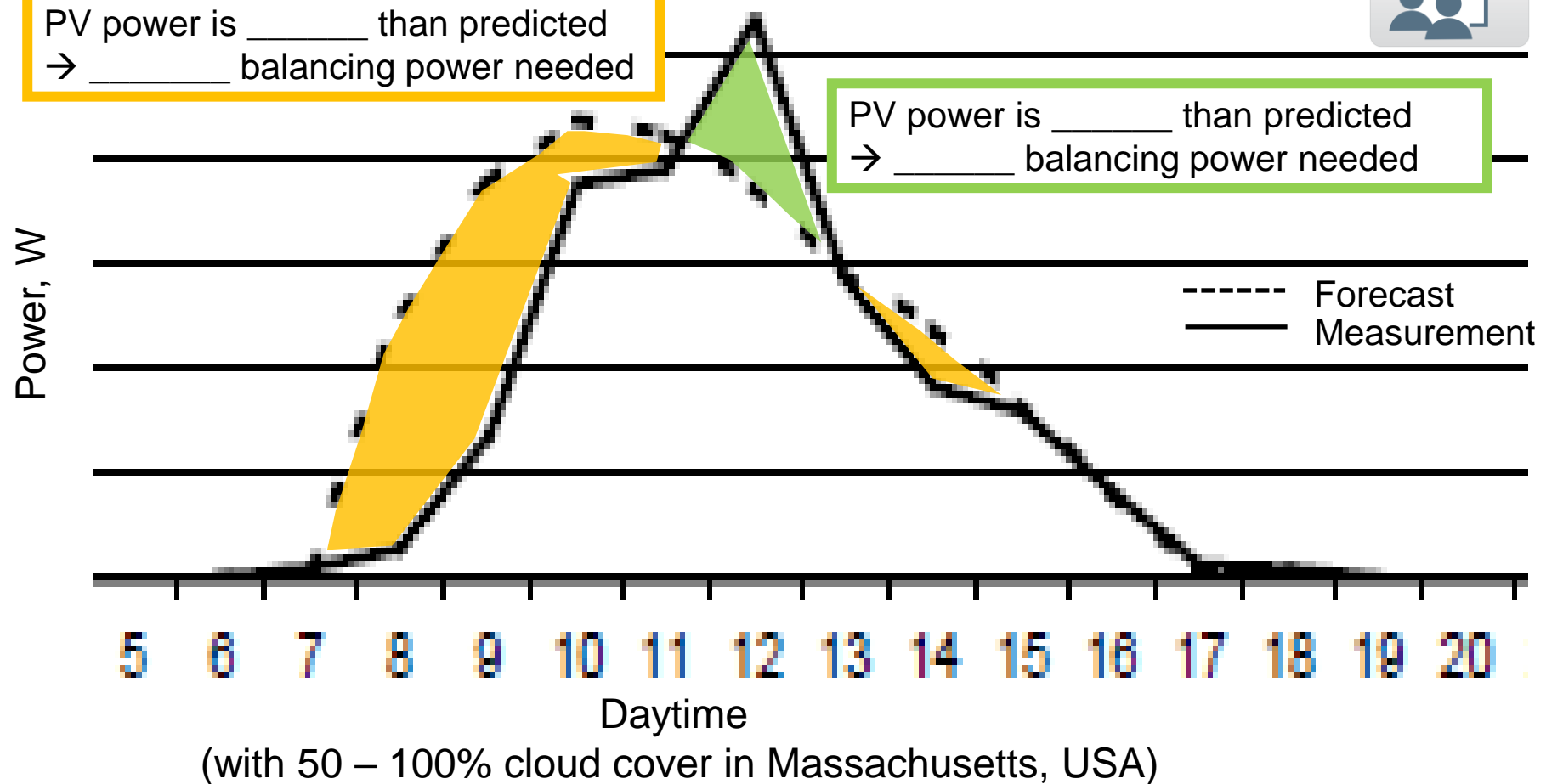
Source: ISET, 2003, supplemented

# Exercise: Which imbalances need positive and negative balancing power

- PV day ahead forecast and online feed-in (measurement)

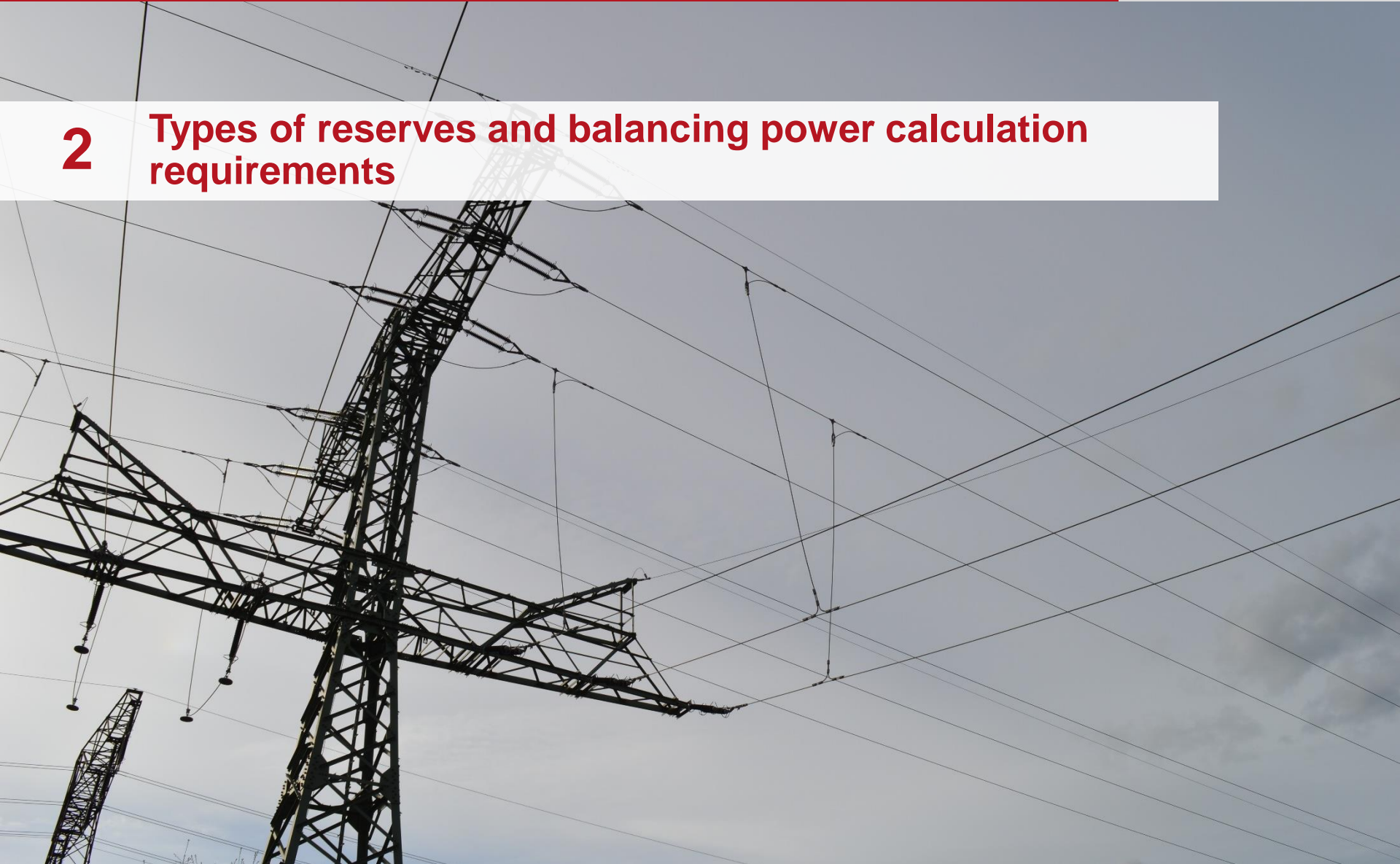
PV power is \_\_\_\_\_ than predicted  
→ \_\_\_\_\_ balancing power needed

PV power is \_\_\_\_\_ than predicted  
→ \_\_\_\_\_ balancing power needed



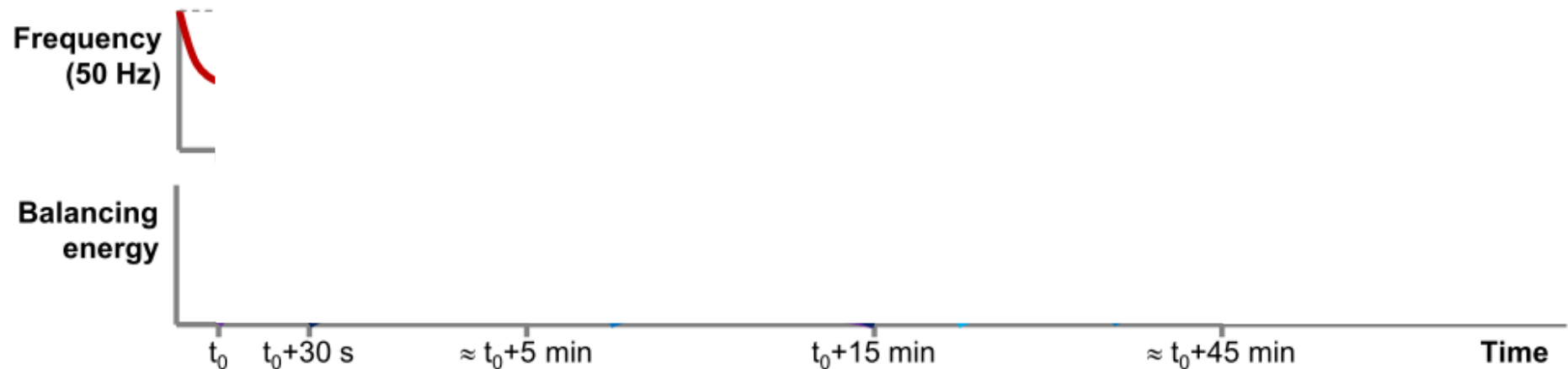


## 2 Types of reserves and balancing power calculation requirements



# Types of reserves (Fast Containment Reserve, Frequency Restoration Reserve and Replacement Reserve)

<b>FCR</b> Automatic activation Up to 15 minutes	<b>aFRR</b> Automatic activation 30s to 15 minutes	<b>mFRR</b> Semi-automatic or manual activation Minimum 15 min.	<b>RR</b> Semi-automatic or manual activation Minimum 15 min.
--	--	--	--



FCR = Fast Containment Reserve

aFRR = automatic Frequency Restoration Reserve

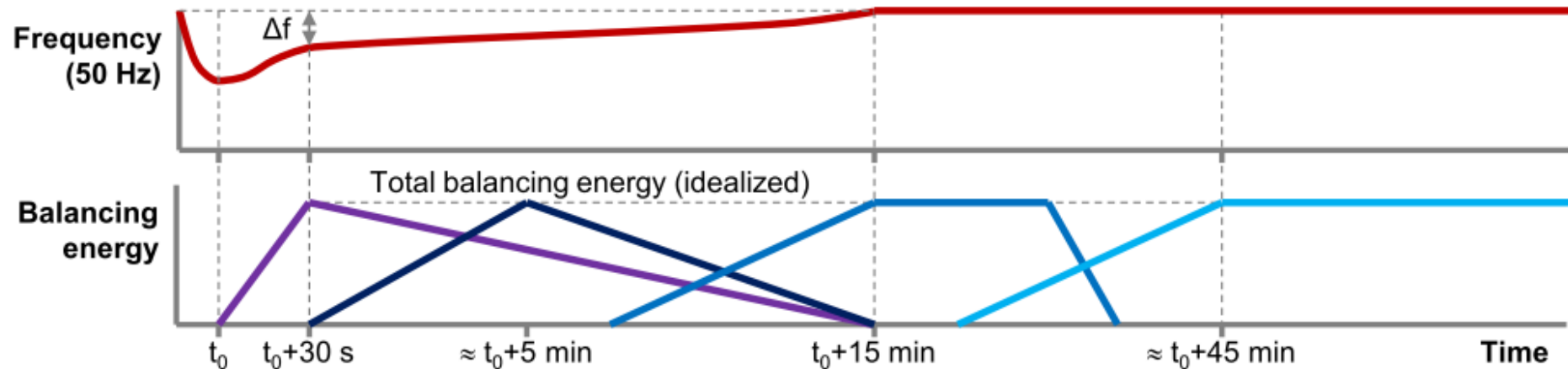
mFRR = manual Frequency Restoration Reserve

RR = Replacement Reserve

Kjell A Barmnes, Convenor WG AS, ENTSO-E and Alexander Dusolt, Market Advisor, ENTSO-E:  
„ELECTRICITY BALANCING GUIDELINE AND IMPLEMENTATION“, page 15, 25 April 2017

# Types of reserves (Fast Containment Reserve, Frequency Restoration Reserve and Replacement Reserve)

<b>FCR</b> Automatic activation Up to 15 minutes	<b>aFRR</b> Automatic activation 30s to 15 minutes	<b>mFRR</b> Semi-automatic or manual activation Minimum 15 min.	<b>RR</b> Semi-automatic or manual activation Minimum 15 min.
--	--	--	--



FCR = Fast Containment Reserve

aFRR = automatic Frequency Restoration Reserve

mFRR = manual Frequency Restoration Reserve

RR = Replacement Reserve

Kjell A Barmnes, Convenor WG AS, ENTSO-E and Alexander Dusolt, Market Advisor, ENTSO-E:  
„ELECTRICITY BALANCING GUIDELINE AND IMPLEMENTATION“, page 15, 25 April 2017

## Balancing power needs – deterministic calculation

- Some system operators calculate the need for primary, secondary or tertiary reserve according to the largest power station in their control zone.
- Example
  - Ability to balance two very large power plant blackouts at the same time
  - Largest power plant capacity ca. 375 MW
  - $\text{Reserve} = 2 \times 375 \text{ MW} = 750 \text{ MW}$
- Disadvantages of this approach are:
  - Over- or undersupply with reserves at certain times is not known
  - Power generation with wind and photovoltaic can not be included in the model

# Types of reserves (Fast Containment Reserve and Frequency Restoration Reserve)

	Primary reserve FCR	Secondary reserve aFRR	Tertiary reserve mFRR
Function			
Activation time			
Responsibility			
Dimensioning			

FCR = Fast Containment Reserve

aFRR = automatic Frequency Restoration Reserve

mFRR = manual Frequency Restoration Reserve

RR = Replacement Reserve

# Types of reserves (Fast Containment Reserve and Frequency Restoration Reserve)

	Primary reserve FCR	Secondary reserve aFRR	Tertiary reserve mFRR
Function	Stopping and stabilising frequency deviations	<ul style="list-style-type: none"> <li>- Return the frequency to its target value</li> <li>- Balance the deviations between control areas</li> </ul>	
Activation time	< 30 s	< 5 min	< 15 min
Responsibility	Decentralized based on a solidarity principle within the synchronous area	Each system operator is technically and economically responsible for its own control area based on the 'user pays principle'	
Dimensioning	Administrative by the transmission code	<b>Probabilistic dimensioning methods</b>	

FCR = Fast Containment Reserve

aFRR = automatic Frequency Restoration Reserve

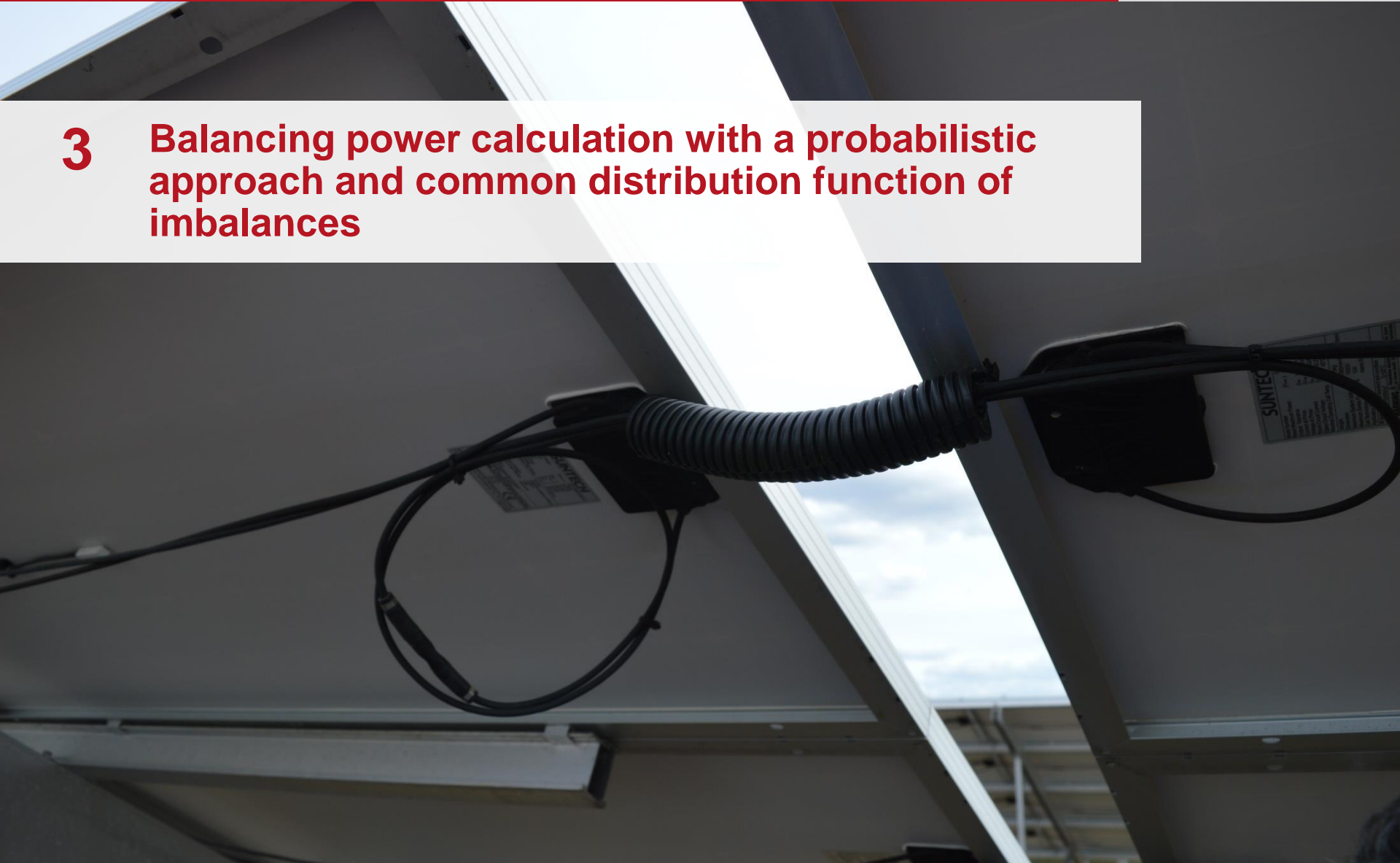
mFRR = manual Frequency Restoration Reserve

RR = Replacement Reserve



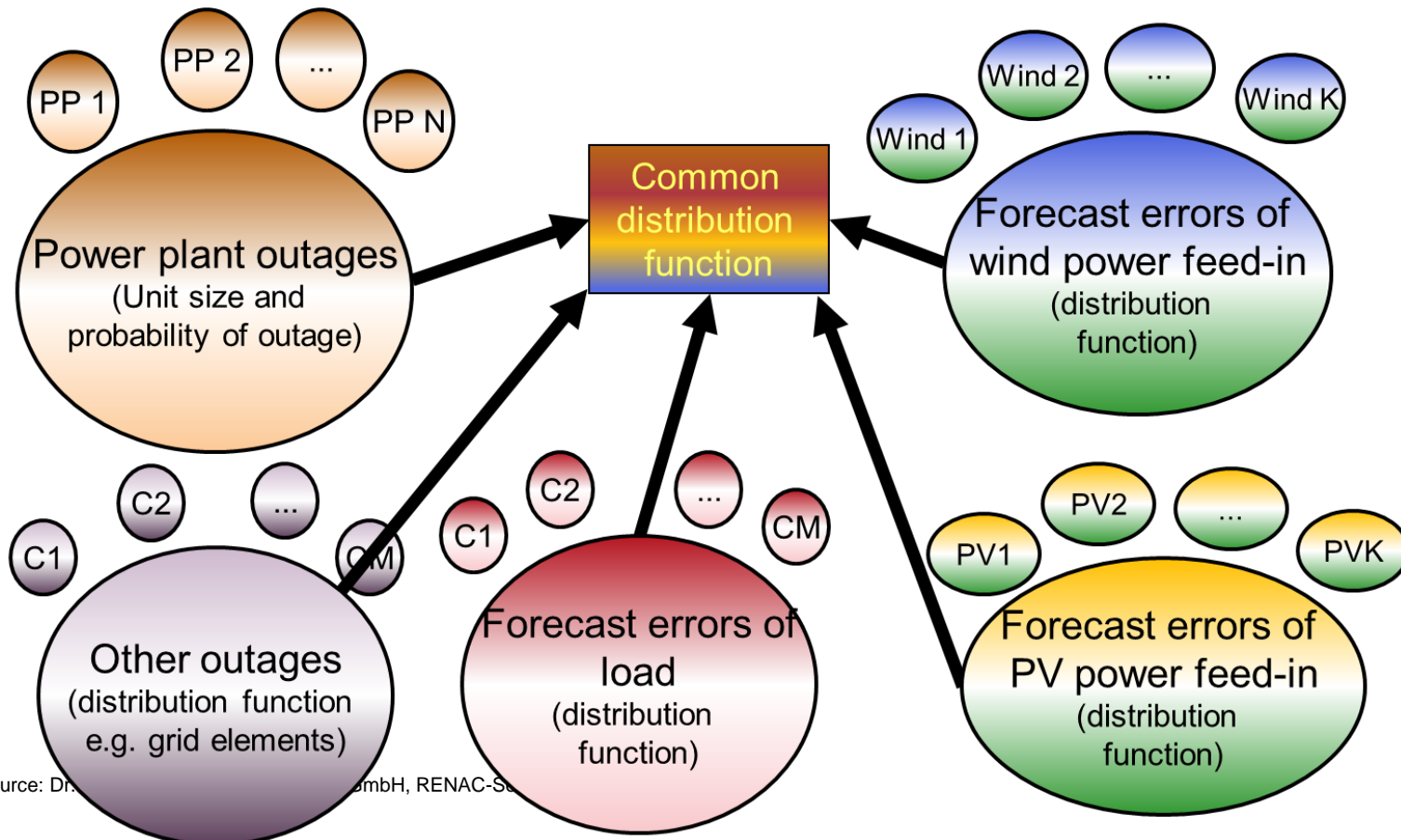
- Probabilistic dimensioning of secondary (aFFR) and tertiary reserves (mFFR) should take into account the following:
  - Short term forecast errors of wind power and solar power
  - Unexpected blackout of power stations and deviation from schedules
  - Load forecast error
  - Gradient of wind power and solar power feed in → residual load gradient
  - Deviation of power stations from planned schedule
  - Outage of transmission and distribution elements, and others
- System reliability level that shall be achieved
- Time specific data for dynamic calculation

### 3 Balancing power calculation with a probabilistic approach and common distribution function of imbalances

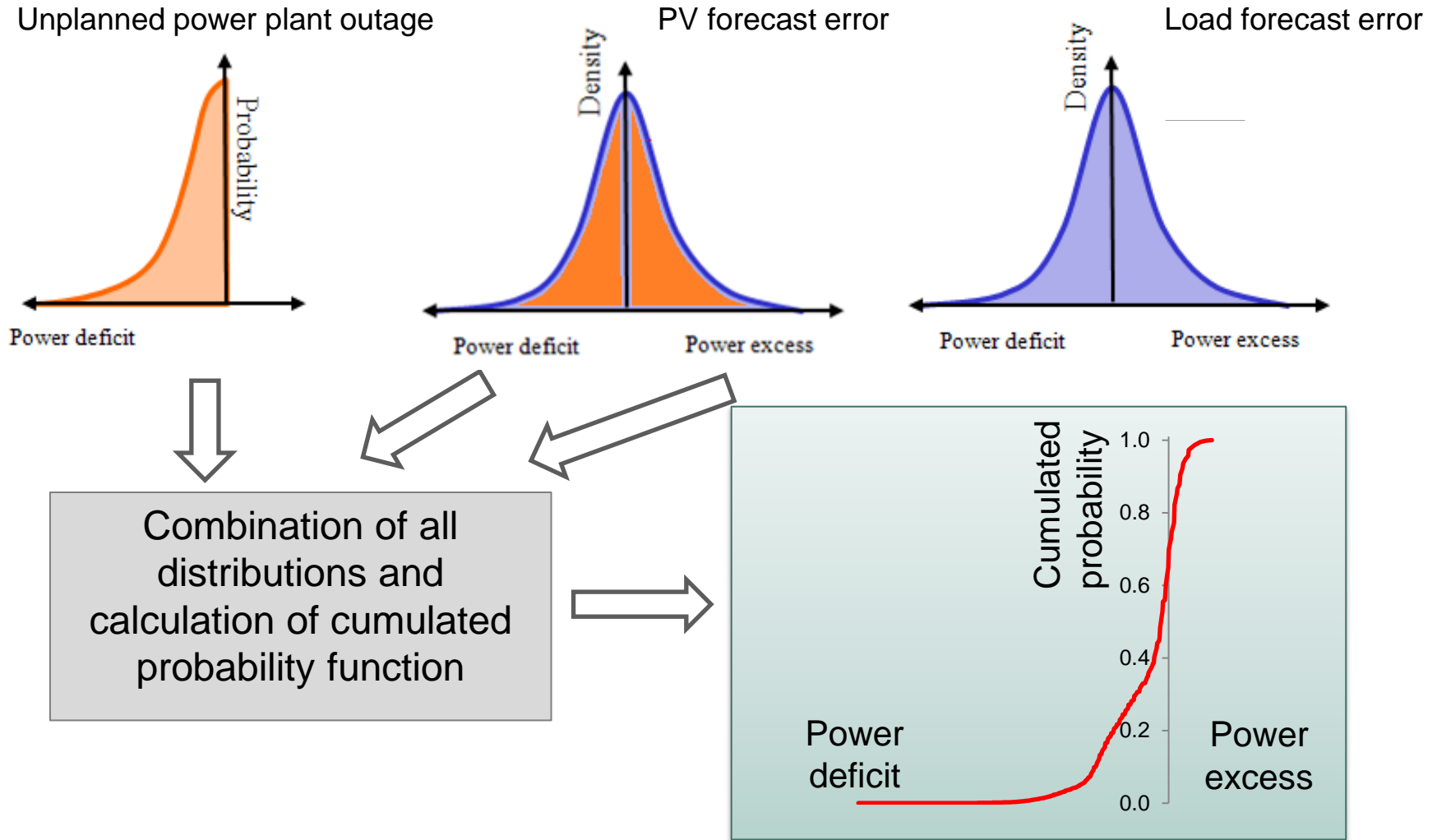


# Common distribution function inputs

- Common distribution function combines different imbalance distributions
- The calculation of the cumulated imbalance distribution needs computational power and adequate algorithms

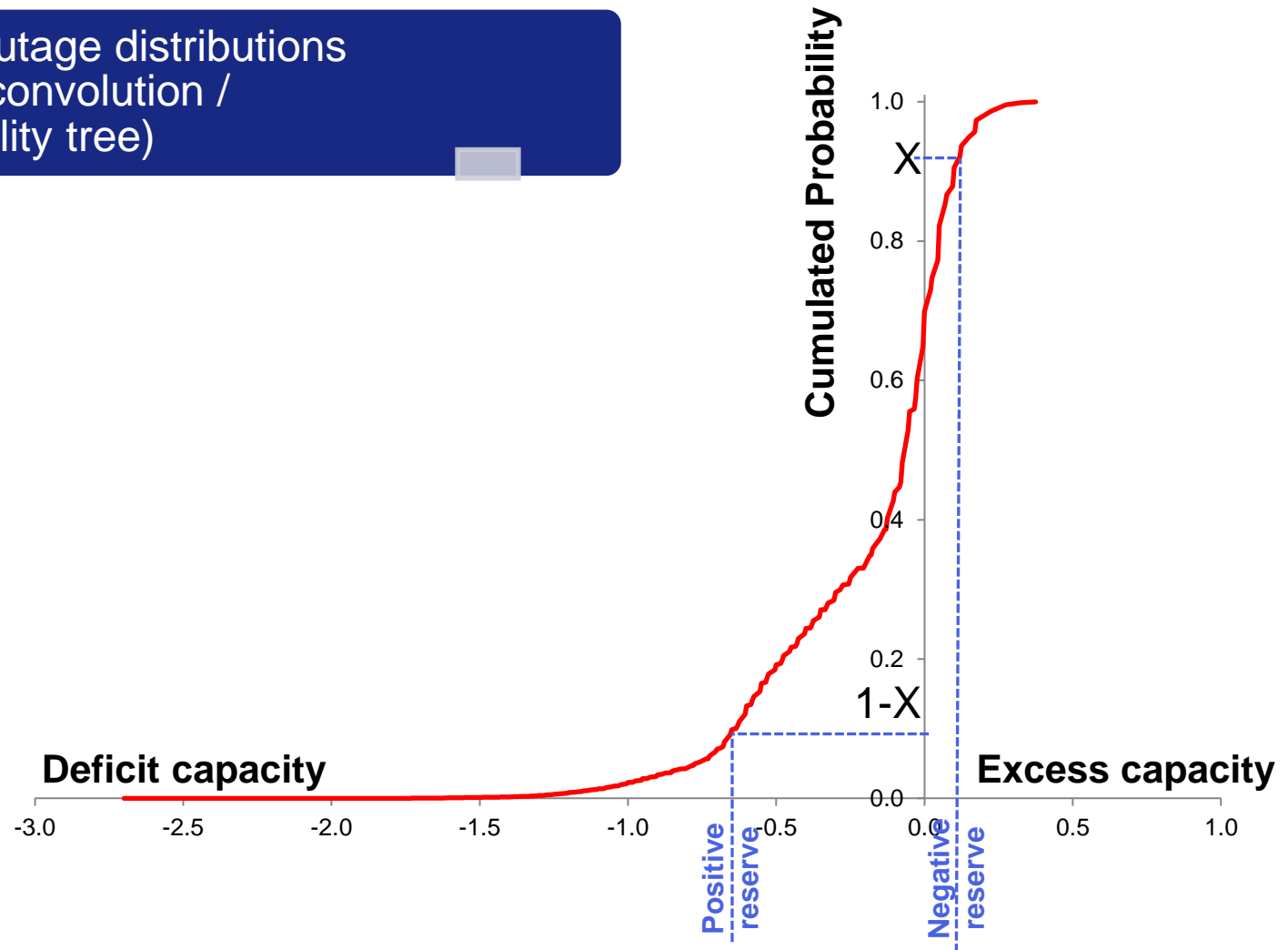


# Overview - probabilistic balancing power calculation



# Steps of probabilistic balancing power calculation methodology with reliability level

Combination of outage distributions  
(recursive convolution /  
probability tree)



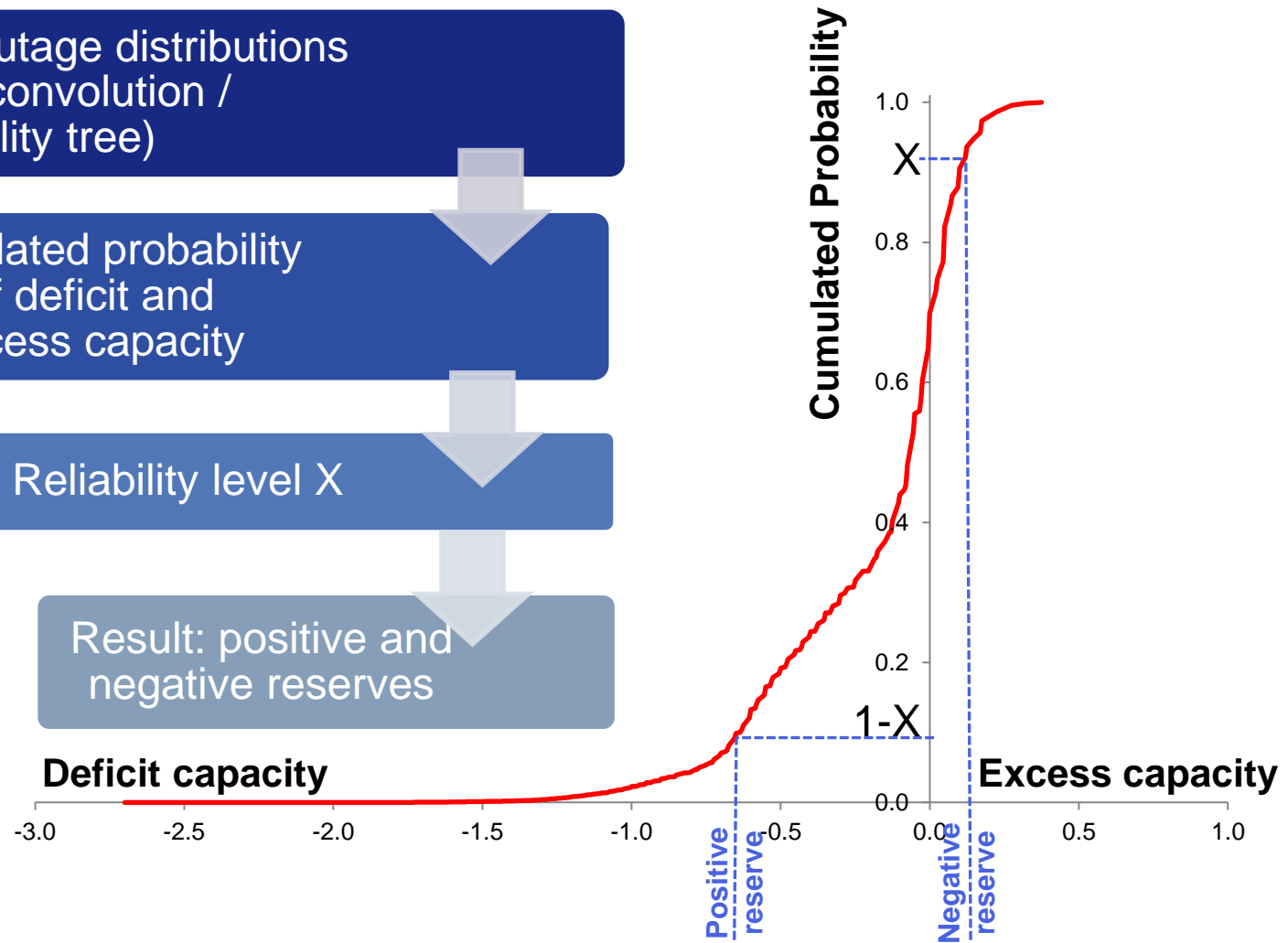
# Steps of probabilistic balancing power calculation methodology with reliability level

Combination of outage distributions  
(recursive convolution /  
probability tree)

Cumulated probability  
of deficit and  
excess capacity

Reliability level  $X$

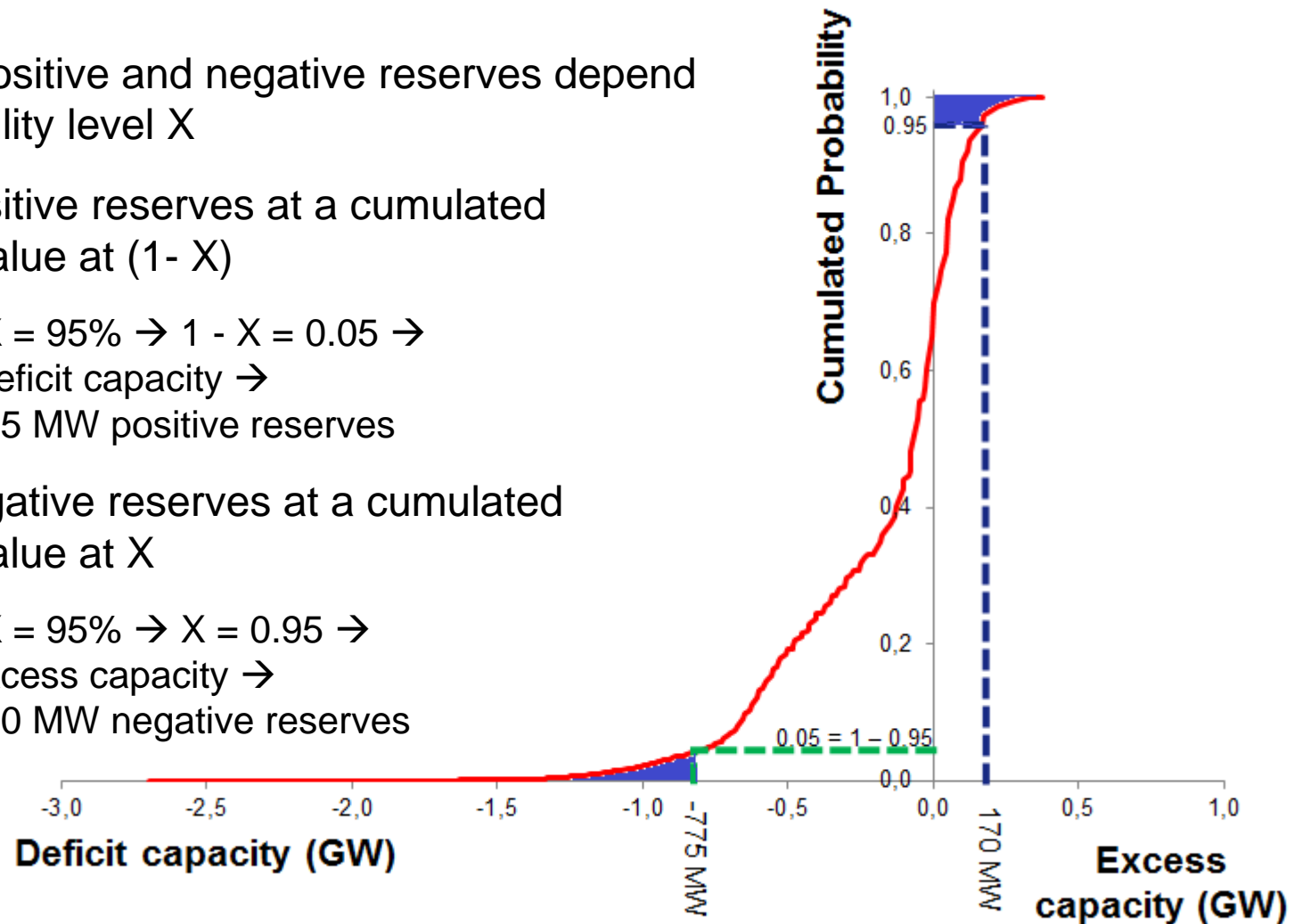
Result: positive and  
negative reserves





# Reserve calculation with cumulated outage and excess capacity, cumulated probability and reliability level

- Amount of positive and negative reserves depend on the reliability level  $X$
- Need for positive reserves at a cumulated probability value at  $(1 - X)$ 
  - **Example:**  $X = 95\% \rightarrow 1 - X = 0.05 \rightarrow$   
-775 MW deficit capacity  $\rightarrow$   
need for 775 MW positive reserves
- Need for negative reserves at a cumulated probability value at  $X$ 
  - **Example:**  $X = 95\% \rightarrow X = 0.95 \rightarrow$   
170 MW excess capacity  $\rightarrow$   
need for 170 MW negative reserves



## 3a Power plant outage distribution

# Definition of the power plant outage distribution

- To understand the properties of the aggregated power plant outages, we need to understand the outage probabilities of each power plant.
- Definition outage probability  
The outage probability of a power plant is defined as the unplanned and undispatchable non-availability of a power plant within a given time frame.
- Outage probability can empirically be found by analyzing the failure statistics of individual power plants.
- Additionally, we need a probability model that transforms the failure statistics into an outage probability.

# Power plant outage model: Markov Process

- A simple outage model of power plants is a two-state homogenous Markov-Process:
  - State A: The power plant is running  $E[T_A]$  is the average time (in hours), a power plant operates without any failure
  - State B: The power plant drops out  $E[T_B]$  is the average time (in hours), a power plant is out of order
- Assumptions:
  - Drop-out rate is independent from time and constant
  - The lifetime of a power plant equals infinity (this is just a mathematical assumption of little relevance)
- Result:
  - The resulting outage probabilities are time independent and can be expressed as a function of operation and outage times

# Power plant outage model: Markov Process

- The probability of a power plant being in operation is  $\Pr\{A\}$
- The probability of being out of order is  $\Pr\{B\}$

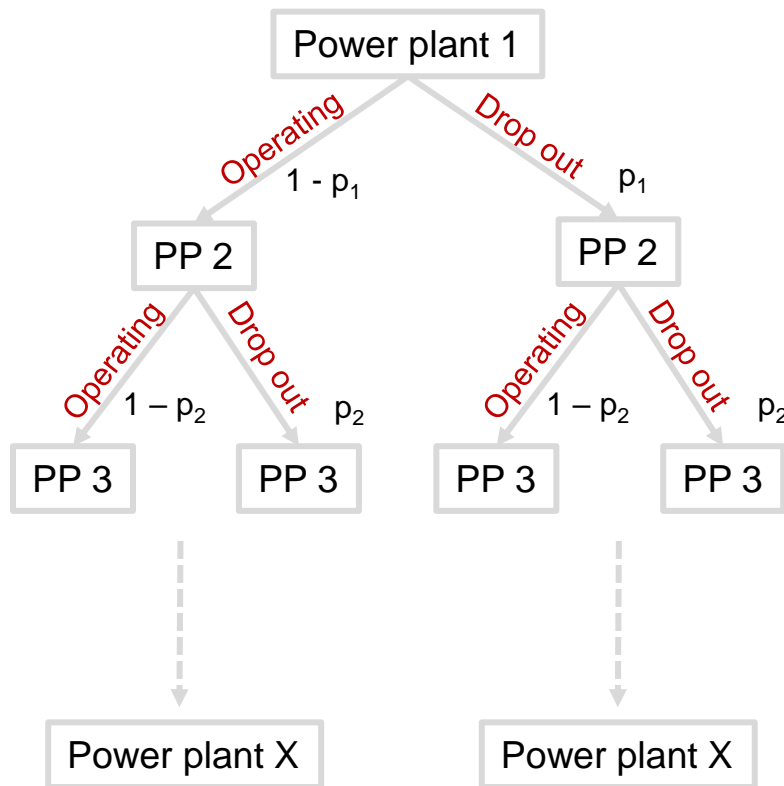
$$\Pr\{A\} = \frac{E[T_A]}{E[T_A] + E[T_B]}$$

$$\Pr\{B\} = \frac{E[T_B]}{E[T_A] + E[T_B]}$$

With:

- Power plant is running  $E[TA]$
- Power plant drops out  $E[TB]$

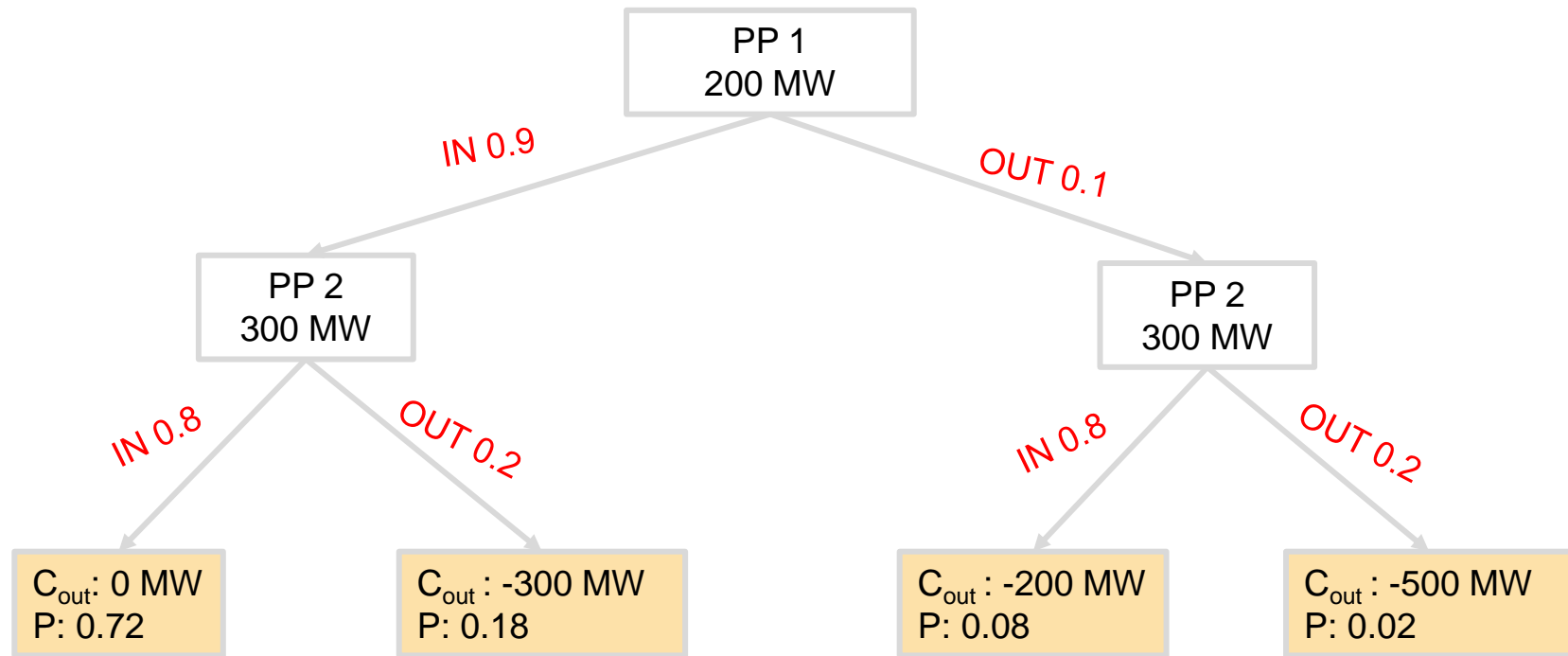
# Power plant outage tree methodology



- All possible combinations of power plant drop-outs are calculated. Each resulting state is characterized by a total failed capacity and a (cumulated) probability of occurrence
- Each power plant  $i$  is characterized by an outage probability  $p_i$  and an installed capacity  $c_i$
- With the probability  $p_i$  the capacity  $c_i$  becomes unavailable
- The tree has  $2^n$  nodes (simplifications in the search process are possible to reduce complexity)
- Large power generation systems need computational power to calculate the tree



# Outage probability tree for two power stations



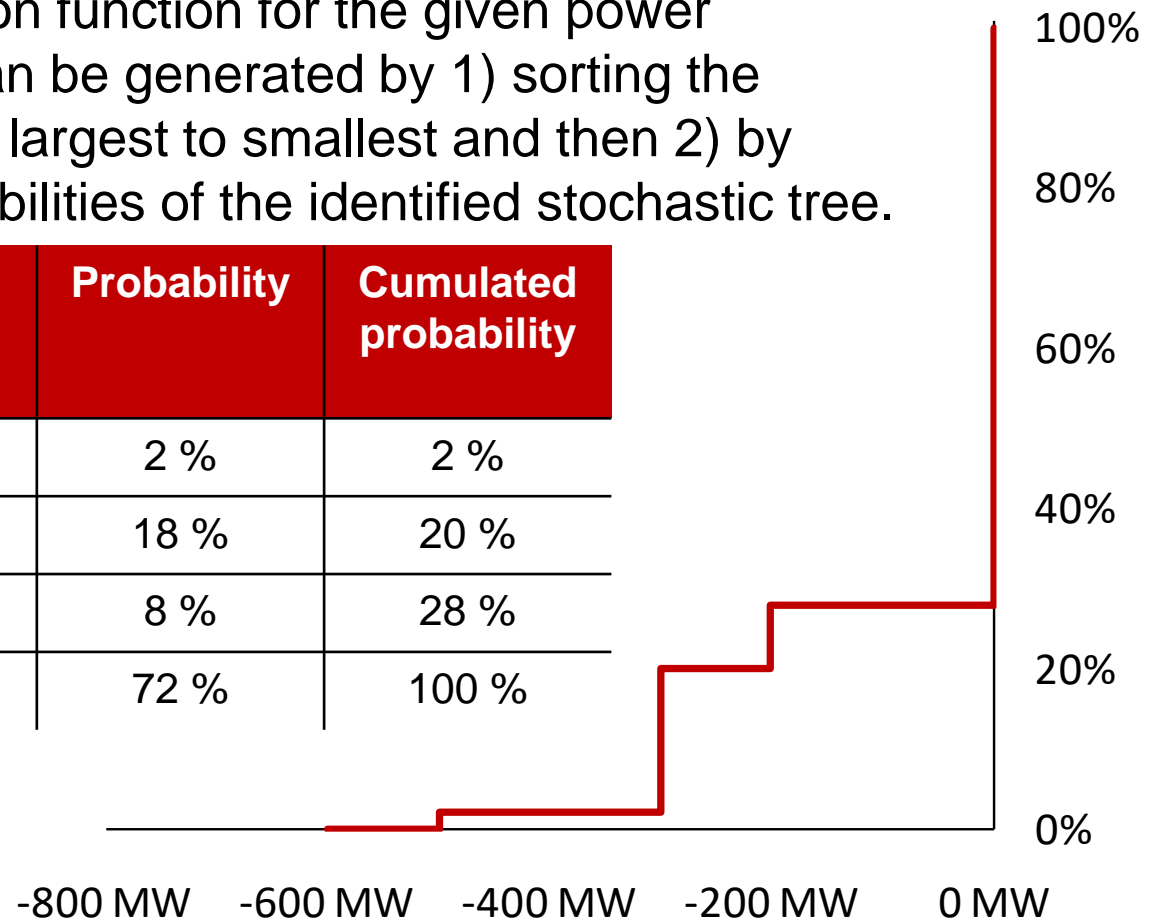
PP = Power plant  
 C<sub>out</sub> = Capacity outage (MW)  
 R = Reserve (MW)

P = Probability (1 = 100%)  
 IN = PP in operation  
 OUT = PP out of operation

# Cumulated outage distribution for two power stations

- The outage distribution function for the given power generation system can be generated by 1) sorting the outage capacity from largest to smallest and then 2) by cumulating the probabilities of the identified stochastic tree.

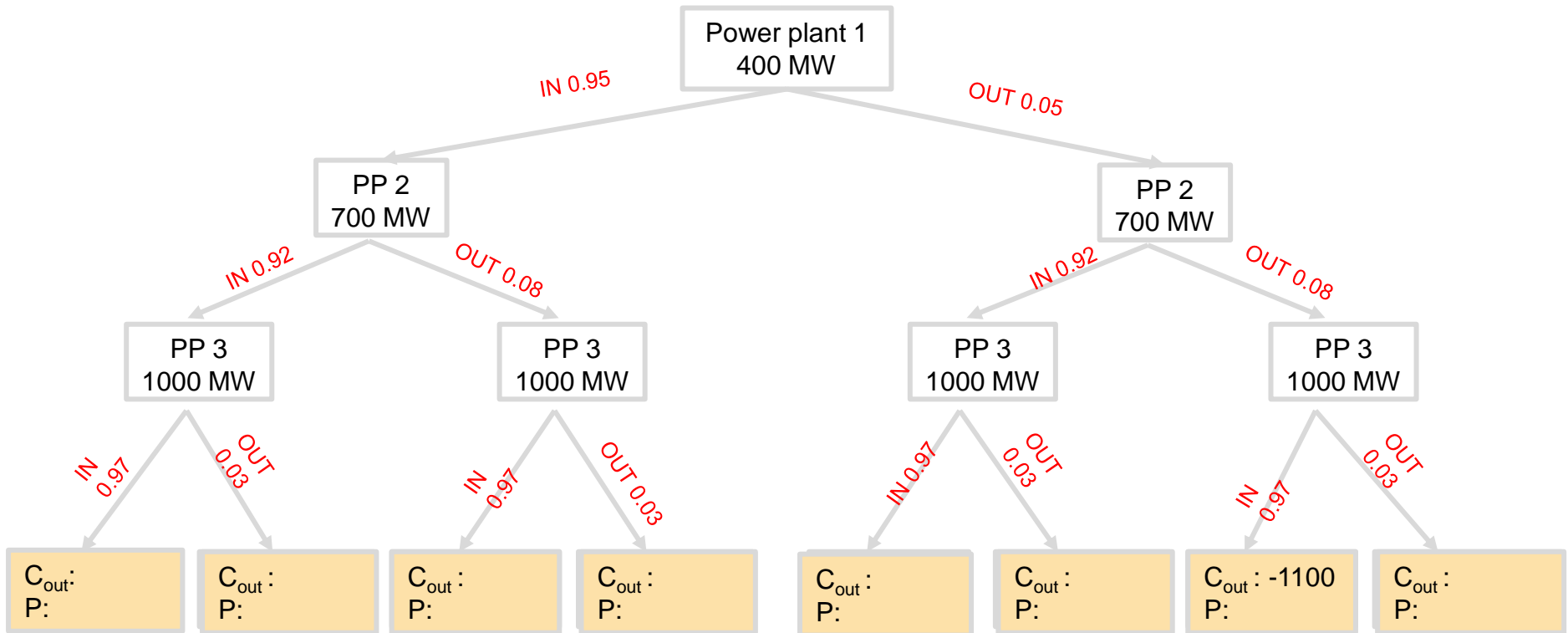
Tree (O=out, I=In)	Sorted outage capacity	Probability	Cumulated probability
OO	- 500 MW	2 %	2 %
IO	- 300 MW	18 %	20 %
OI	- 200 MW	8 %	28 %
II	0 MW	72 %	100 %



# Exercise: Cumulated outage distribution graph for three power stations

- A fictitious small country has only three thermal power plants. The three plants have the following capacities and outage probabilities:
  - Power plant 1: 400 MW, outage probability 5%
  - Power plant 2: 700 MW, outage probability 8%
  - Power plant 3: 1000 MW, outage probability 3%
- To do's:
  - Draw the probability tree for the three plants on paper
  - Calculate probabilities and outage capacity for each node in the tree and write the probabilities on the respective place in the probability tree
  - Sort the outage capacity from largest to smallest
  - Calculate the cumulated outage probability for each outage combination

# Exercise: Probability tree



PP = Power plant  
 C<sub>out</sub> = Capacity outage (MW)  
 R = Reserve (MW)

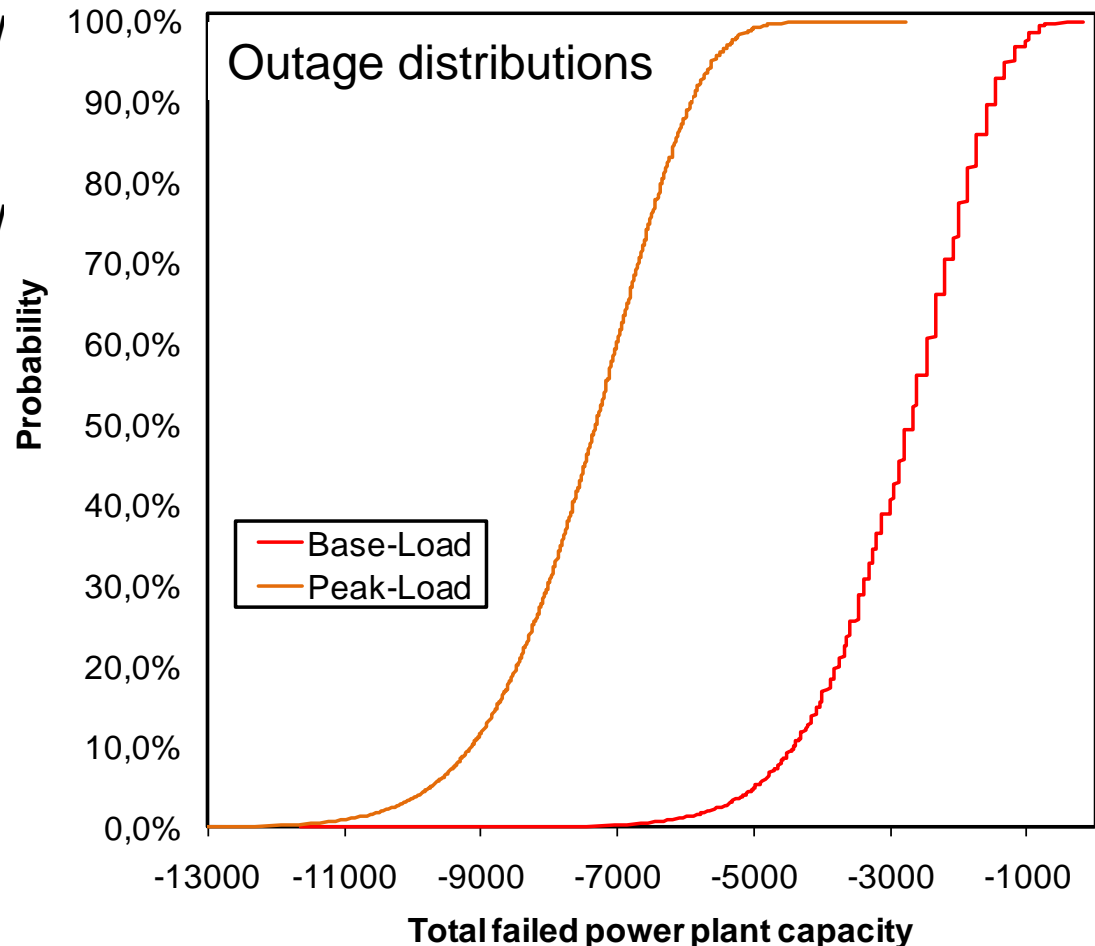
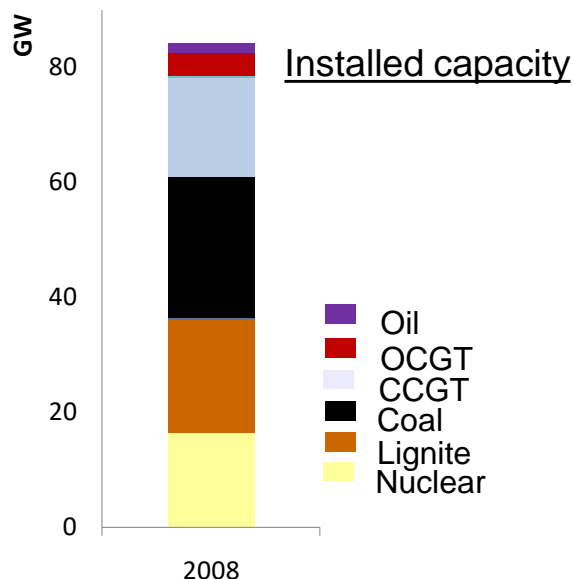
P = Probability (1 = 100%)  
 IN = PP in operation  
 OUT = PP out of operation

## Exercise: Cumulated outage probability

Power station i = in, o = out	Probability	Outage capacity (MW)	Cumulated outage probability
ooo			
ioo			
oio			
ooi			
iiio			
ioi			
oii			
iii			

# Example: Outage distributions for a system with 40 GW off peak and 80 GW peak load

- Power plant data set consists of 723 generation units.
- Case 'base-load': ~ 40 GW of power plants connected to cover the base load
- Case 'peak-load': ~ 80 GW of power plants connected to cover peak load

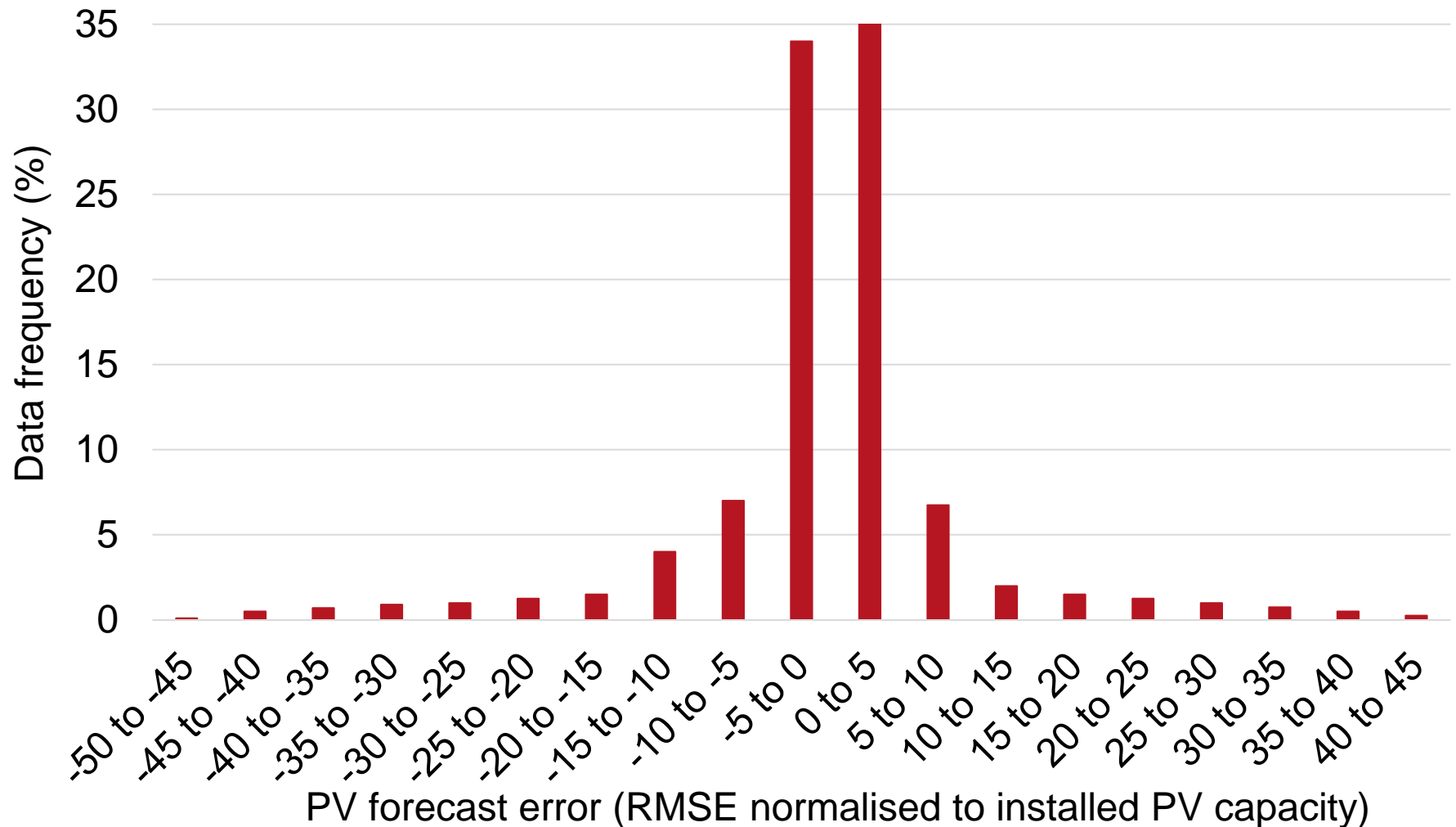




## PV Forecast error distribution



# PV forecast error distribution example



Source: internal data

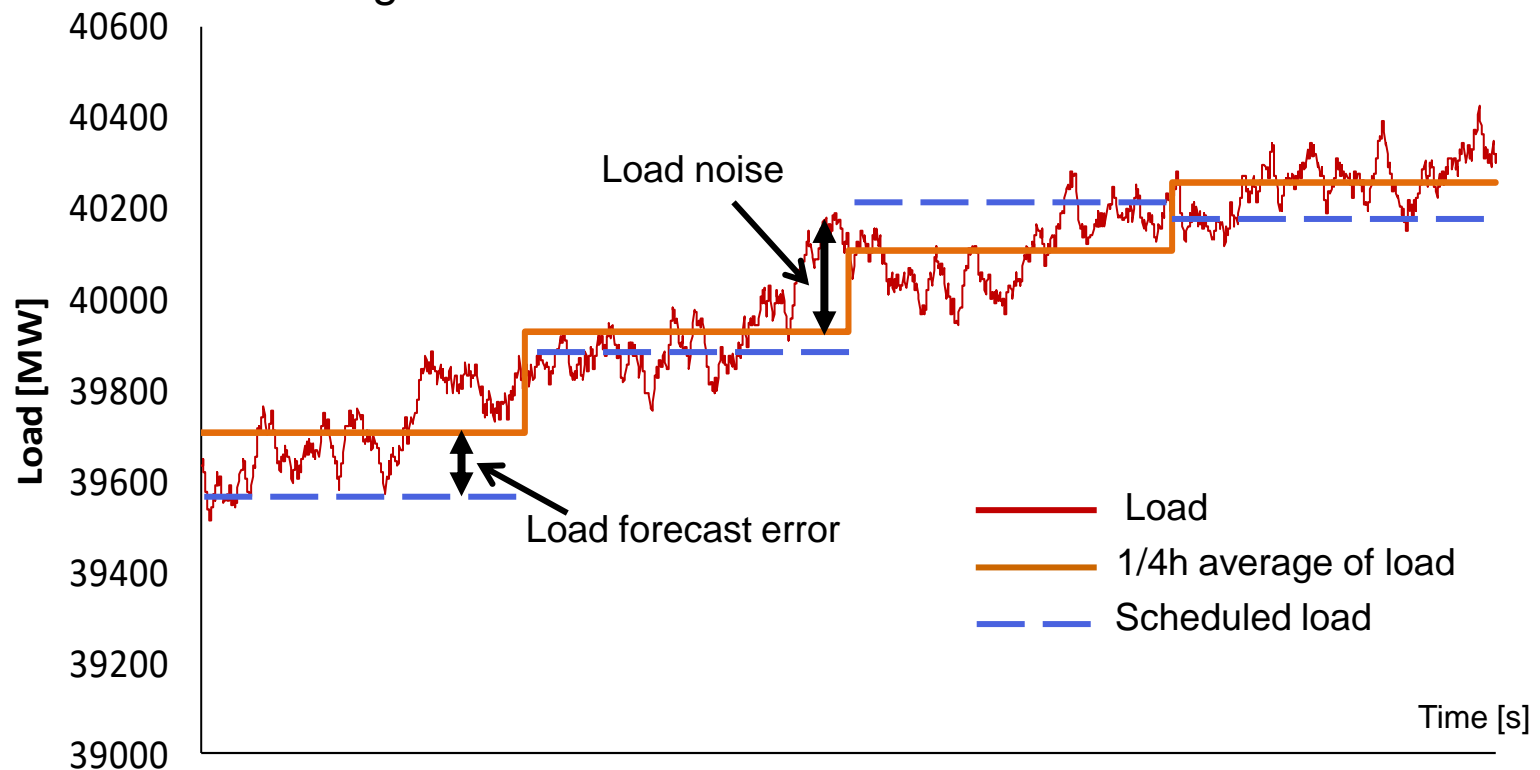


## 3b Load forecast error distribution

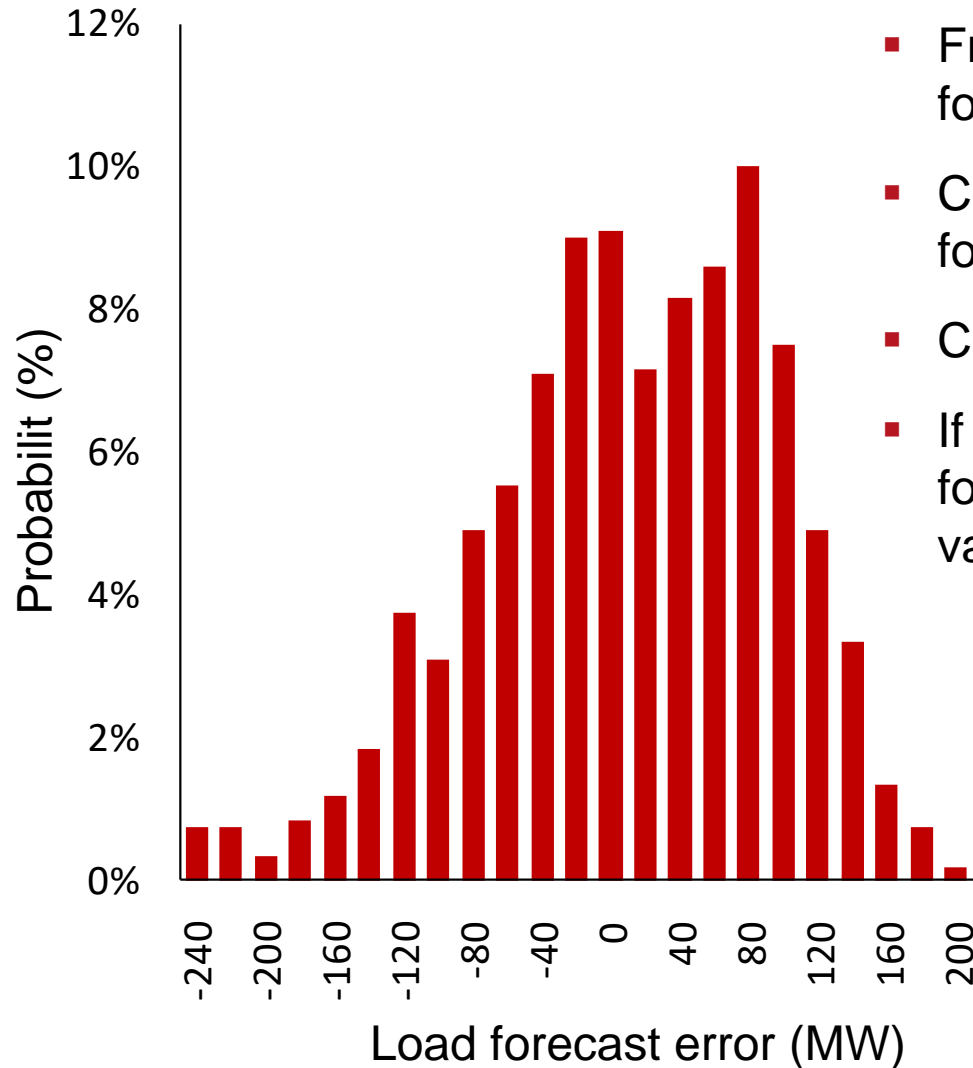


# Load forecast errors and load noise

- **Load noise** is the difference between the load at a specific moment and the average of load in the respective 15 minute interval
- **Load forecast error** is the difference between the scheduled load and the average of measured load in a 15 minute interval



# Load forecast error distribution example



- From empirical values, build the load forecast distribution function
- Count the frequency of each load forecast within a given range
- Construct the empirical histogram
- If desired, an analytical form can be found by interpolating the empirical values

Minus = Underestimation of load  
 Plus = Overestimation of load



## 4 **Balancing power calculation with probabilistic approach - exercises with excel tool**



# Balancing power calculation Excel tool

- Input:
  - Power station outage distribution (up to 100 power stations)
  - Wind forecast error distribution
  - PV Forecast error distribution
  - Load forecast error distribution and further distributions
  - Reliability level (power outage / excess power)
- Output:
  - Power outage → positive reserve
  - Excess power → negative reserve

1. Enter the power plants and their outage probabilities into the sheet “Powerplants”
2. Calculate the cumulated power plant outage distribution
3. Enter the first additional distribution (e.g. wind forecast error) into the sheet “Input 2” (or whatever its name is)
4. Calculate the cumulated imbalance distribution by adding the first additional distribution
5. Perform step 3 and 4 for up to four other additional distributions (e.g. PV or load forecast errors) by using the sheets “Input 3” to “Input 6”

**Analysis&Result**

Powerplants

Input 2

Input 3

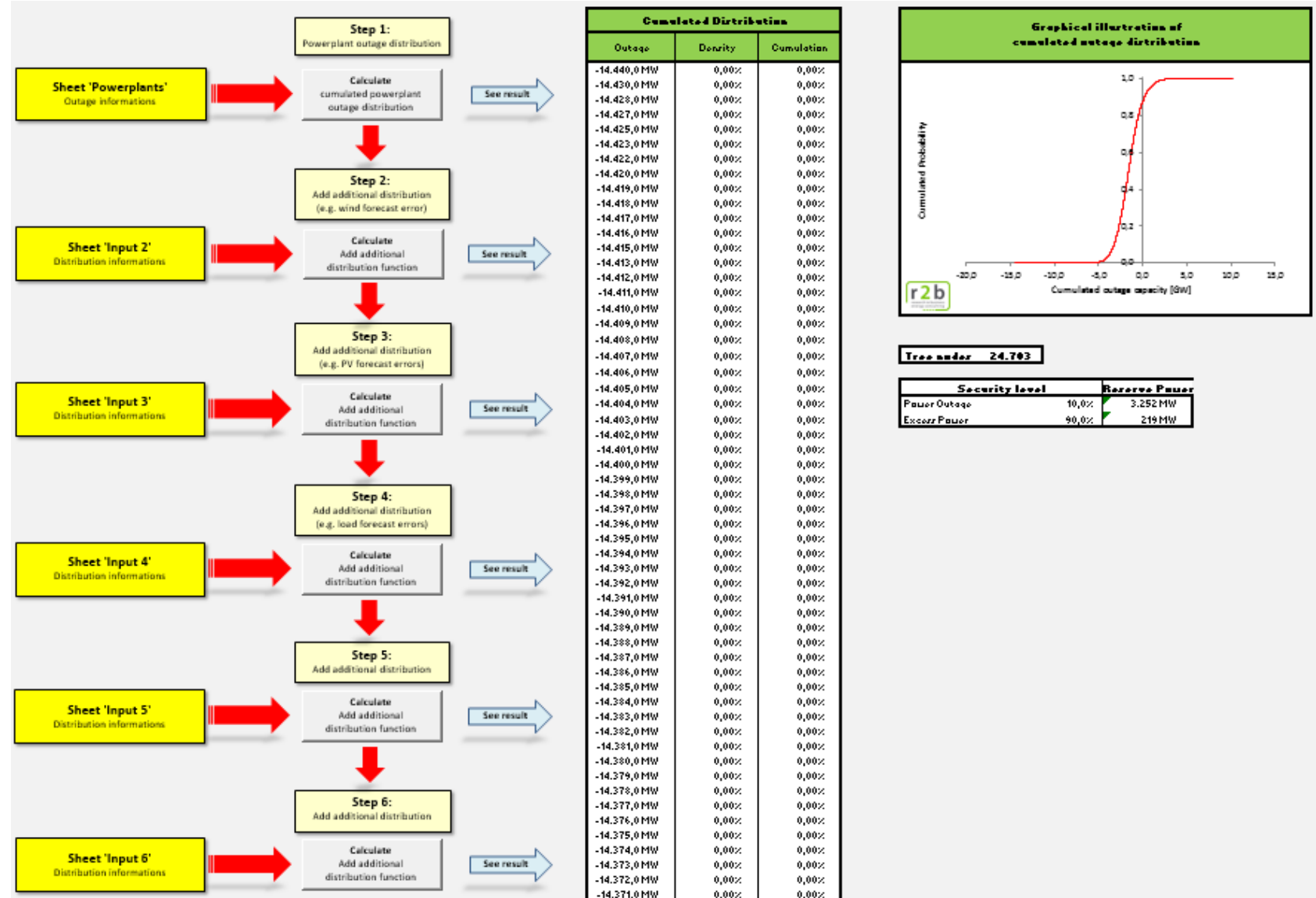
Input 4

Input 5

Input 6

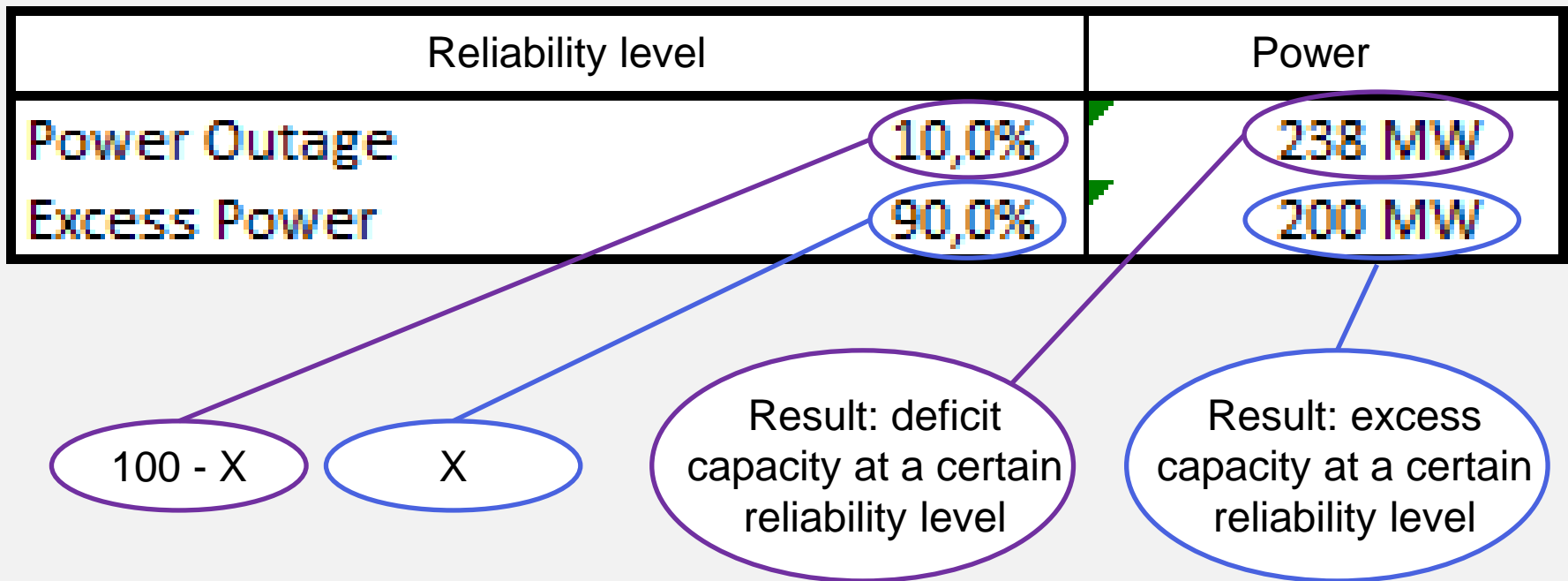


# Main sheet - control area (left) and result area (right)



# Balancing power calculation Excel tool

- Input:
  - Reliability level X:
    - Power Outage Probability =  $100 - X$
    - Excess power probability =  $X$



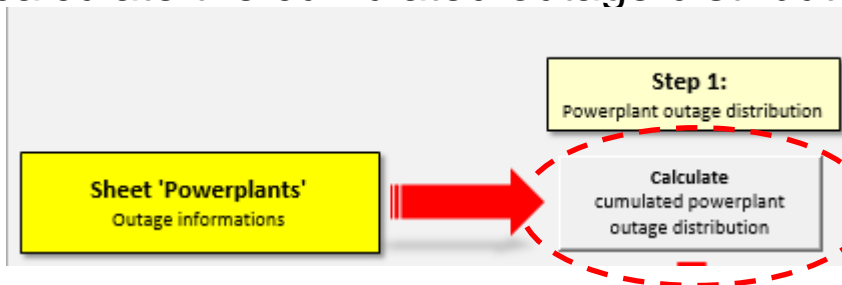
## Step 1: Adding power plants

- Enter power plant data (up to 100 power plants)

Powerplant data		
Nr.	Capacity	Probability of unforced Outage
1		
2		
3		
4		
5		
6		
7		
8		
9		



- When you have finished adding your power plants, click the yellow button to return to the sheet “Analysis&Results”. Press “Calculate” button to calculate the cumulated outage distribution.



## Sign of load and wind /PV forecast error



- Load forecast error:
  - **Positive** values mean that less power is consumed than predicted and this means that additional power station capacity is available. (→ add positive numbers to excel tool)
  - **Negative** values mean that more power is consumed than predicted and the power generation capacity is less than predicted. (→ add negative numbers to excel tool)
- Wind and PV forecast error:
  - **Positive** values mean that wind / PV power is overestimated and this means that less power is available than predicted. This is comparable to an unexpected black-out of a power station (→ add negative numbers to excel tool)
  - **Negative** values mean that wind / PV power is underestimated and this means that more power is available than predicted. (→ add positive numbers to excel tool)

## 4a Exercises 1 and 2



## Balancing power exercises 1 and 2



- Calculate the positive and negative balancing power needs for a reliability level of 95% and 99.9%.
- Exercise 1:
  - Assume 20 power stations with a capacity of 100 MW to 450 MW. The next two slides show data of the power stations' capacity and outage probability.
  - Assume a PV forecast error and load forecast error according to the following slides.
- Exercise 2:
  - Change the power stations. Assume only 10 power stations. Please combine power station 1+2, 3+4...19+20. Note that total capacity remains unchanged. All probabilities are the same, as the individual plants.
  - Keep the PV forecast and load forecast data constant (same as in exercise 1)



# Exercises 1 and 2: Probabilistic balancing power calculation (power stations, PV and load)



95% Reliability	Positive balancing power	Negative balancing power
System 1 with 20 medium size power stations	MW	MW
System 2 with 10 large power stations	MW	MW

99.9% Reliability	Positive balancing power	Negative balancing power
System 1 with 20 medium size power stations	MW	MW
System 2 with 10 large power stations	MW	MW

## Exercise 1 - input data for 20 power stations



### Power station input data:

- thermal power plant 1: 150 MW, outage probability 5%
- thermal power plant 2: 150 MW, outage probability 5%
- thermal power plant 3: 200 MW, outage probability 8%
- thermal power plant 4: 200 MW, outage probability 8%
- thermal power plant 5: 200 MW, outage probability 3%
- thermal power plant 6: 200 MW, outage probability 3%
- thermal power plant 7: 300 MW, outage probability 12%
- thermal power plant 8: 300 MW, outage probability 12%
- thermal power plant 9: 300 MW, outage probability 8%
- thermal power plant 10: 300 MW, outage probability 8%



## Exercise 1 - input data for 20 power stations



### Power station Input data:

- thermal power plant 11: 100 MW, outage probability 5%
- thermal power plant 12: 100 MW, outage probability 5%
- thermal power plant 13: 300 MW, outage probability 8%
- thermal power plant 14: 300 MW, outage probability 8%
- thermal power plant 15: 350 MW, outage probability 3%
- thermal power plant 16: 350 MW, outage probability 3%
- thermal power plant 17: 400 MW, outage probability 12%
- thermal power plant 18: 400 MW, outage probability 12%
- thermal power plant 19: 450 MW, outage probability 8%
- thermal power plant 20: 450 MW, outage probability 8%

## Exercise 2 - input data for 10 power station

### Input data:

- thermal power plant 1: 300 MW, outage probability 5%
- thermal power plant 2: 400 MW, outage probability 8%
- thermal power plant 3: 400 MW, outage probability 3%
- thermal power plant 4: 600 MW, outage probability 12%
- thermal power plant 5: 600 MW, outage probability 8%
- thermal power plant 6: 200 MW, outage probability 5%
- thermal power plant 7: 600 MW, outage probability 8%
- thermal power plant 8: 700 MW, outage probability 3%
- thermal power plant 9: 800 MW, outage probability 12%
- thermal power plant 10: 900 MW, outage probability 8%



## Exercise 1 and 2 - PV forecast error distribution



### Input data:

- Overestimation 250 MW: 3%
  - Overestimation 150 MW: 7%
  - Overestimation 45 MW: 16%
  - Overestimation 25 MW: 26%
- 
- Underestimation 30 MW: 24%
  - Underestimation 50 MW: 14%
  - Underestimation 160 MW: 7%
  - Underestimation 275 MW: 3%

Add negative MW-values to the excel tool

Add positive MW-values to the excel tool

## Exercise 1 and 2 - Load forecast error distribution

### Input data:

- Overestimation 125 MW: 7%
  - Overestimation 75 MW: 15%
  - Overestimation 25 MW: 28%
- 
- Underestimation 25 MW: 28%
  - Underestimation 75 MW: 15%
  - Underestimation 125 MW: 7%



Add positive MW-values to the excel tool

Add negative MW-values to the excel tool

## 4b Exercise 3



## Exercise 3: Balancing power calculation (only power outages and load forecast error – no wind and no PV)



- Calculate the positive and negative balancing power needs for a reliability level of 95% and 99,9% for a power system with **3318 MW installed thermal capacity and two different outage probabilities but no PV generation.**
  - Scenario A
    - Assume power station outages for **scenario A**. The next two slides shows data of the power stations' capacity and outage probability A and B.
    - Use load forecast error data (next slide)
  - Scenario B
    - Assume power station outages for **scenario B**. The next two slides shows data of the power stations' capacity and outage probability A and B.
    - Use the same load forecast error as in scenario A.



## Exercise 3: Probabilistic balancing power calculation (power stations and load error)



95% Reliability	Positive balancing power	Negative balancing power
System with scenario A power stations (without wind and PV)	MW	MW
System with scenario B power stations (without wind and PV)	MW	MW

99.9% Reliability	Positive balancing power	Negative balancing power
System with scenario A power stations (without wind and PV)	MW	MW
System with scenario B power stations (without wind and PV)	MW	MW

## Exercise 3: Power station outage assumption (scenario A and B)

Capacity (MW)	A-Outage probability (%)	B-Outage probability (%)
385.0 MW	5	1
319.0 MW	5	1
359.0 MW	5	1
105.0 MW	5	1
114.0 MW	5	1
114.0 MW	5	1
128.0 MW	5	1
132.0 MW	5	1
52.0 MW	5	1
100.0 MW	5	1
25.0 MW	5	1
100.0 MW	5	1
25.0 MW	5	1

Capacity (MW)	A-Outage probability (%)	B-Outage probability (%)
152.0 MW	5	1
70.0 MW	5	1
20.0 MW	5	1
49.0 MW	5	1
100.0 MW	5	1
100.0 MW	5	1
100.0 MW	5	1
90.0 MW	5	1
100.0 MW	5	1
35.0 MW	5	1
30.0 MW	5	1
50.0 MW	5	1
40.0 MW	5	1



## Exercise 3: Power station outage assumption (scenario A and B)

Capacity (MW)	A-Outage probability (%)	B-Outage probability (%)
60.0 MW	5	1
100.0 MW	5	1
50.0 MW	5	1
17.0 MW	5	1
180.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
15.0 MW	5	1
25.0 MW	5	1
<b>3318</b>		

## Exercise 3: Load forecast error distribution

Average of Error bins (MW)	Percent of the year in the respective bin
475	0.01%
425	0.06%
375	0.10%
325	0.16%
275	0.27%
225	1.13%
175	4.06%
125	11.59%
75	19.45%
25	23.87%

Average of Error bins (MW)	Percent of time in the respective bin
-25	17.59%
-75	11.00%
-125	5.76%
-175	2.67%
-225	1.08%
-275	0.51%
-325	0.50%
-375	0.10%
-425	0.06%
-475	0.03%

## 4c Exercise 4



## Exercise 4: Balancing power calculation (power outages + load + wind + PV forecast error)



- Calculate the positive and negative balancing power needs for a reliability level of 95% and 99.9% for a power system with **3318 MW installed thermal capacity and two different outage probabilities.**
- Scenario A
  - Assume power station outages for **scenario A**. The next two slides show data of the power stations' capacity and outage probability A and B.
  - Use load forecast error, wind forecast error and PV forecast error data (next slide)
- Scenario B
  - Assume power station outages for **scenario B**. The next two slides show data of the power stations' capacity and outage probability A and B.
  - Use the same load forecast error, wind forecast error and PV forecast error data as in scenario A.

## Exercise 4: Balancing power calculation (power outages + load + wind + PV forecast error)



95% Reliability	Positive balancing power	Negative balancing power
System with scenario A power stations	MW	MW
System with scenario B power stations	MW	MW

99,9% Reliability	Positive balancing power	Negative balancing power
System with scenario A power stations	MW	MW
System with scenario B power stations	MW	MW

## Exercise 4: Power station outage assumption (scenario A and B)

Capacity (MW)	A-Outage probability (%)	B-Outage probability (%)
385.0 MW	5	1
319.0 MW	5	1
359.0 MW	5	1
105.0 MW	5	1
114.0 MW	5	1
114.0 MW	5	1
128.0 MW	5	1
132.0 MW	5	1
52.0 MW	5	1
100.0 MW	5	1
25.0 MW	5	1
100.0 MW	5	1
25.0 MW	5	1

Capacity (MW)	A-Outage probability (%)	B-Outage probability (%)
152.0 MW	5	1
70.0 MW	5	1
20.0 MW	5	1
49.0 MW	5	1
100.0 MW	5	1
100.0 MW	5	1
100.0 MW	5	1
90.0 MW	5	1
100.0 MW	5	1
35.0 MW	5	1
30.0 MW	5	1
50.0 MW	5	1
40.0 MW	5	1

## Exercise 4: Power station outage assumption (scenario A and B)

Capacity (MW)	A-Outage probability (%)	B-Outage probability (%)
60.0 MW	5	1
100.0 MW	5	1
50.0 MW	5	1
17.0 MW	5	1
180.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
17.0 MW	5	1
15.0 MW	5	1
25.0 MW	5	1
<b>3318</b>		



## Exercise 4: Load forecast error distribution

Average of Error bins (MW)	Percent of the year in the respective bin
475	0.01%
425	0.06%
375	0.10%
325	0.16%
275	0.27%
225	1.13%
175	4.06%
125	11.59%
75	19.45%
25	23.87%

Average of Error bins (MW)	Percent of time in the respective bin
-25	17.59%
-75	11.00%
-125	5.76%
-175	2.67%
-225	1.08%
-275	0.51%
-325	0.50%
-375	0.10%
-425	0.06%
-475	0.03%

Positive error = Underestimation of load → add positive numbers to excel tool

Negative error = Overestimation of load → add negative numbers to excel tool



## Exercise 4: Wind forecast error distribution

Average of Error bins (MW)	Percent of the year in the respective bin
130	0.01%
110	0.05%
90	0.76%
70	4.60%
50	11.35%
30	17.60%
10	27.31%

Average of Error bins (MW)	Percent of the year in the respective bin
-10	25.18%
-30	9.01%
-50	3.17%
-70	0.81%
-90	0.13%
-110	0.02%

Positive error = Overestimation of wind power → add negative numbers to excel tool  
 Negative error = Underestimation of wind power → add positive numbers to excel tool

Source: Load and load forecast error data Yeulis Rivas <[yrvivas@cne.gob.do](mailto:yrvivas@cne.gob.do)>, mail from 7 May 2019; own calculation

## Exercise 4: PV forecast error distribution

Average of Error bins (MW)	Percent of the year in the respective bin
45	0.02%
40	0.17%
35	0.87%
30	2.17%
25	2.82%
20	8.05%
15	16.45%
10	26.69%
5	22.82%

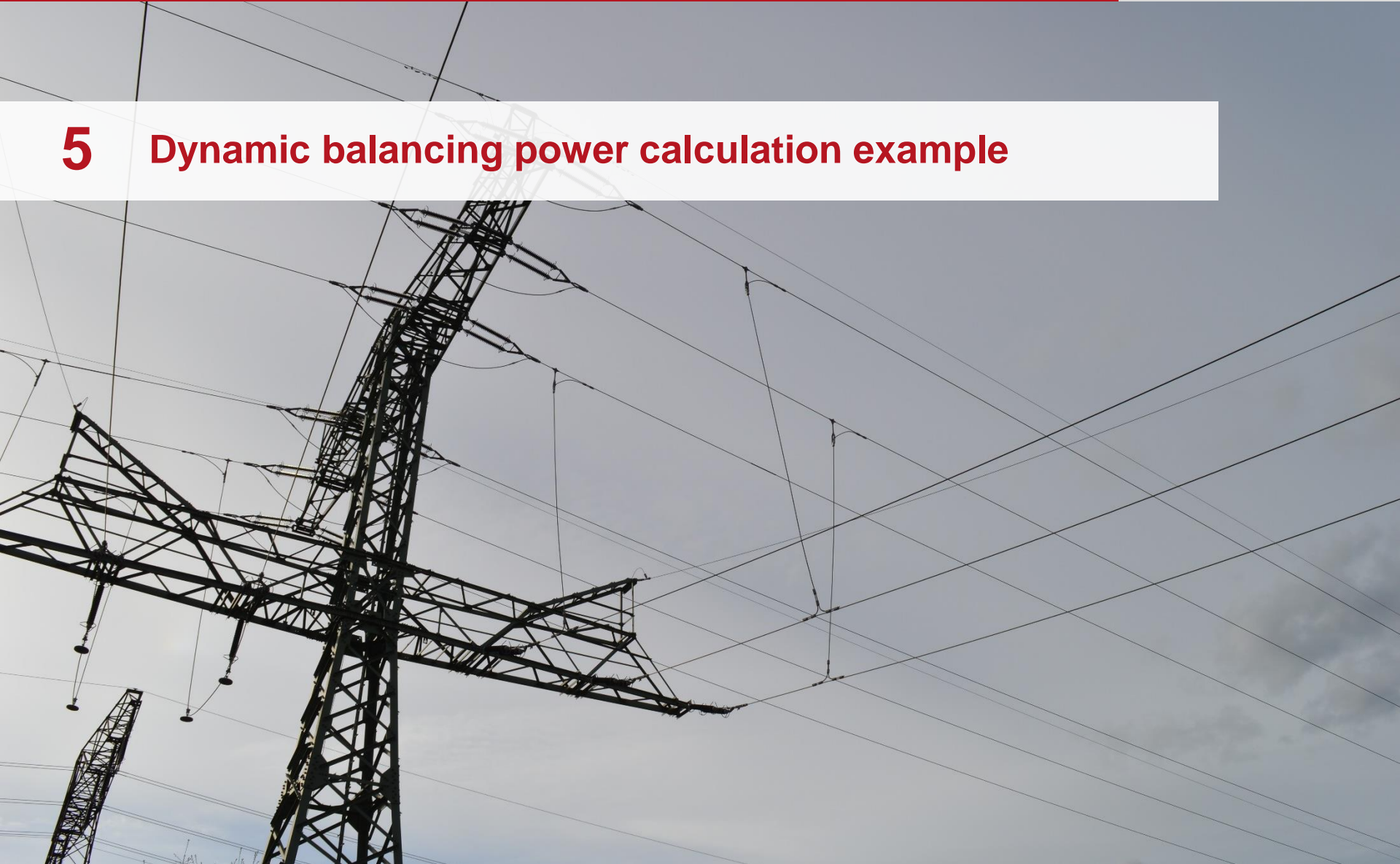
Average of Error bins (MW)	Percent of the year in the respective bin
-5	10.61%
-10	4.92%
-15	2.36%
-20	1.09%
-25	0.55%
-30	0.17%
-35	0.10%
-40	0.12%
-45	0.02%

Positive error = Overestimation of PV power → add negative numbers to excel tool

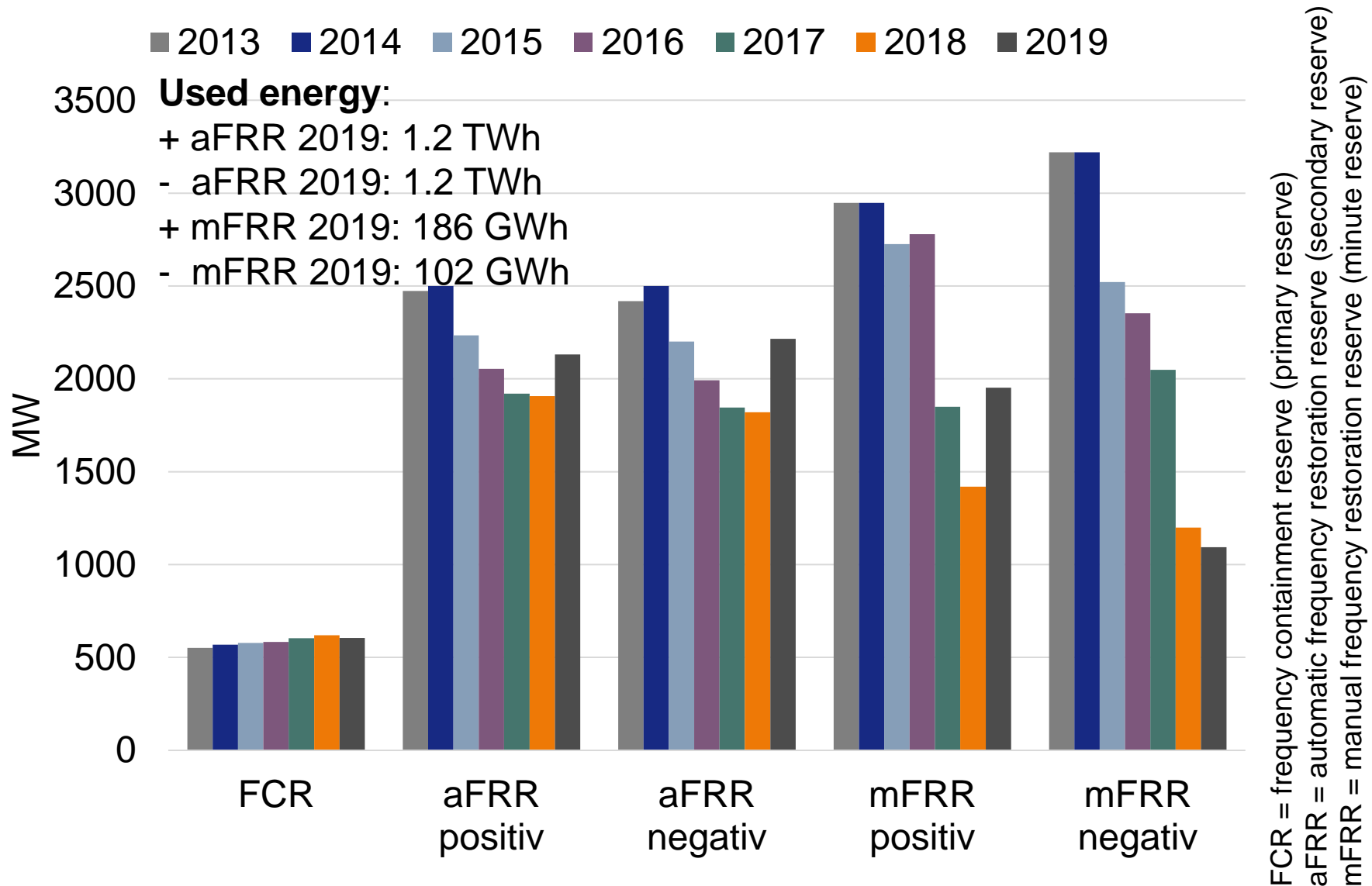
Negative error = Underestimation of PV power → add positive umbers to excel tool

Source: Load and load forecast error data Yeulis Rivas <[yrvivas@cne.gob.do](mailto:yrvivas@cne.gob.do)>, mail from 7 May 2019; own calculation

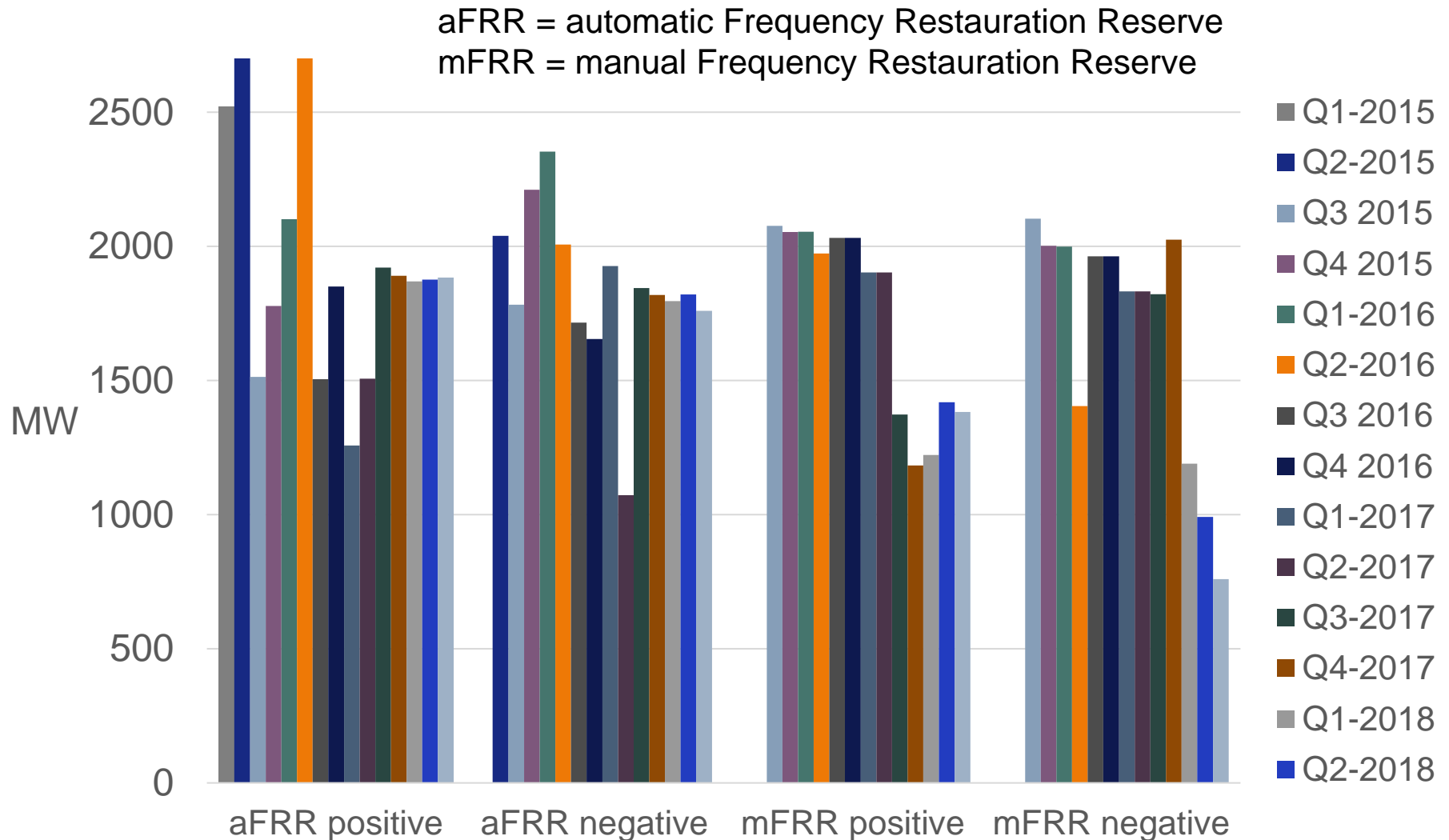
## 5 Dynamic balancing power calculation example



# Development of balancing power call for tender in Germany from 2013 to 2017 (MW, maximum per year)



# Dynamic balancing power tendering in Germany from 2015 - 2018 (aFRR and mFRR)



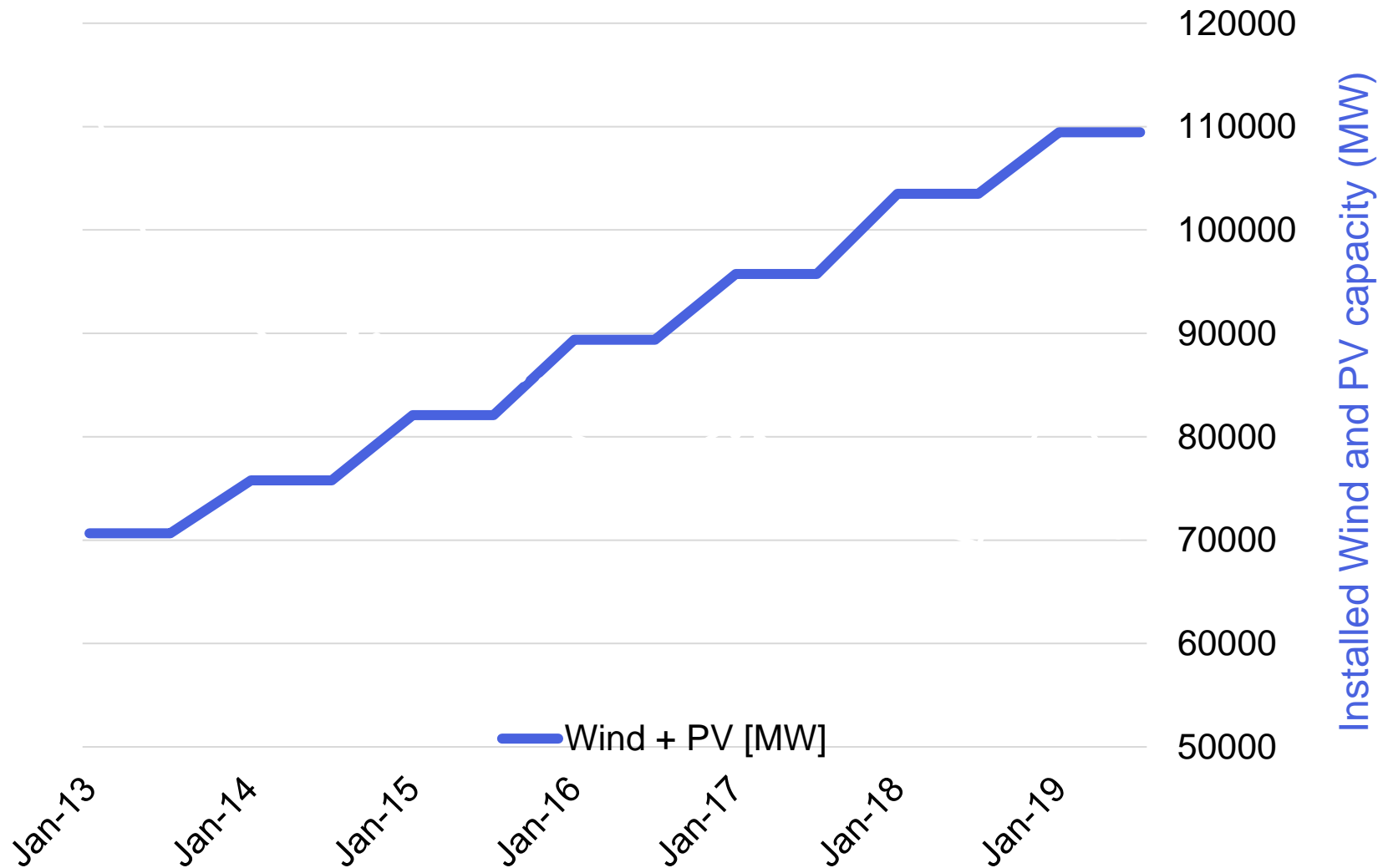
<https://www.regelleistung.net>, 29.8.2018

	Method 1	Method 2
<b>Germany</b>	Until 2019	Since 2020
<b>Calculation cycles</b>	Every three months	Weekly (daily possible)
<b>Data</b>	Identical quarters of previous years	Cluster of representative times
<b>Amount of balancing power</b>	Smaller than deterministic methods	Smaller than method 1

Case study result for one year for 4 TSOs in Germany:

Method	Average negative (MW)	Average positive (MW)	Max negative (MW)	Max positive (MW)	Min negative (MW)	Min positive (MW)
1	-2616	3040	-3010	3295	-2120	2518
2	-2472	2819	-3056	3162	-1558	1924

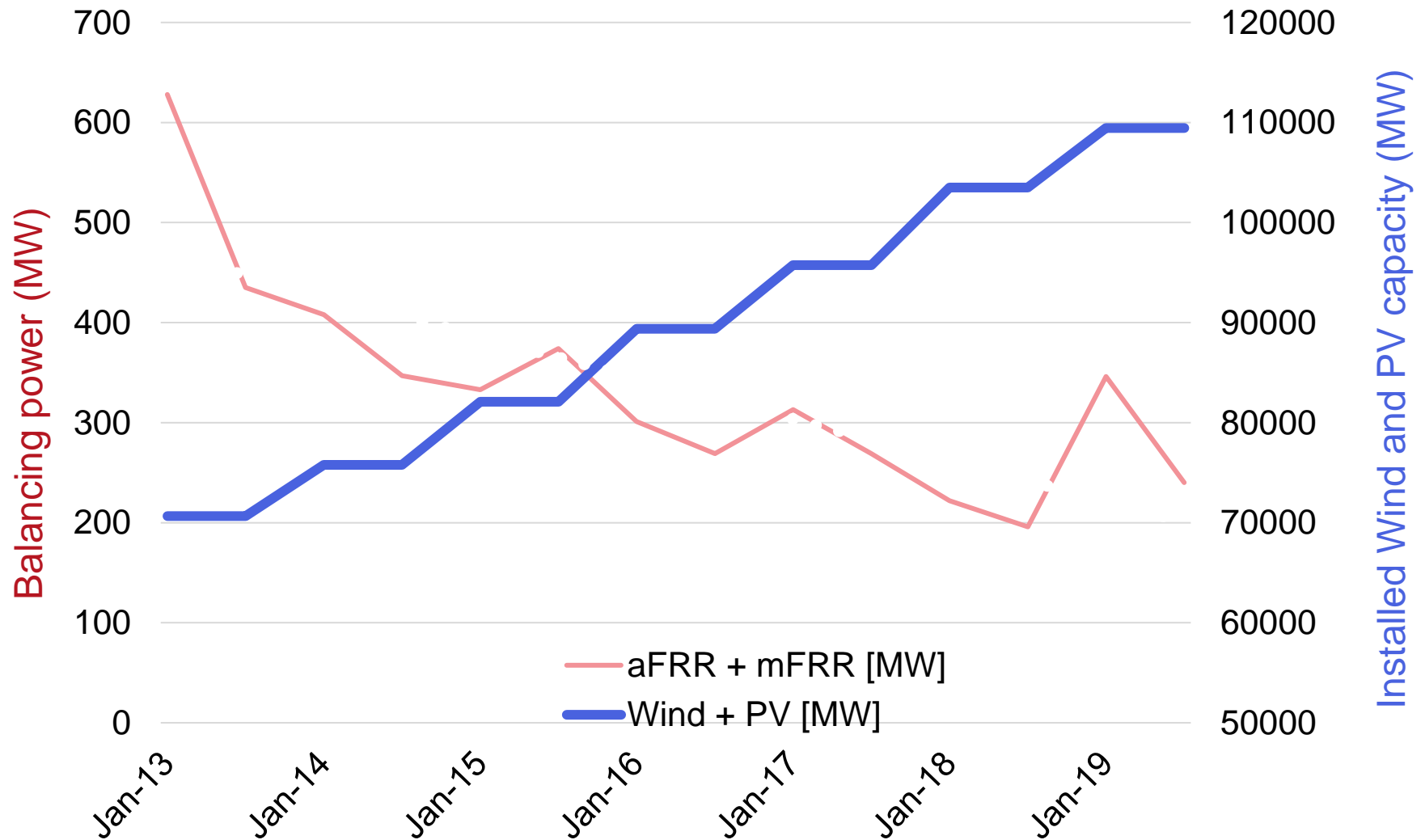
# Development of installed wind & PV capacity and used amount of balancing power (aFRR + mFRR) in Germany



Source: BnetzA „Monitoringbericht 2020“ and BMWi „Zeitreihen zur Entwicklung der erneuerbaren Energien in Deutschland“

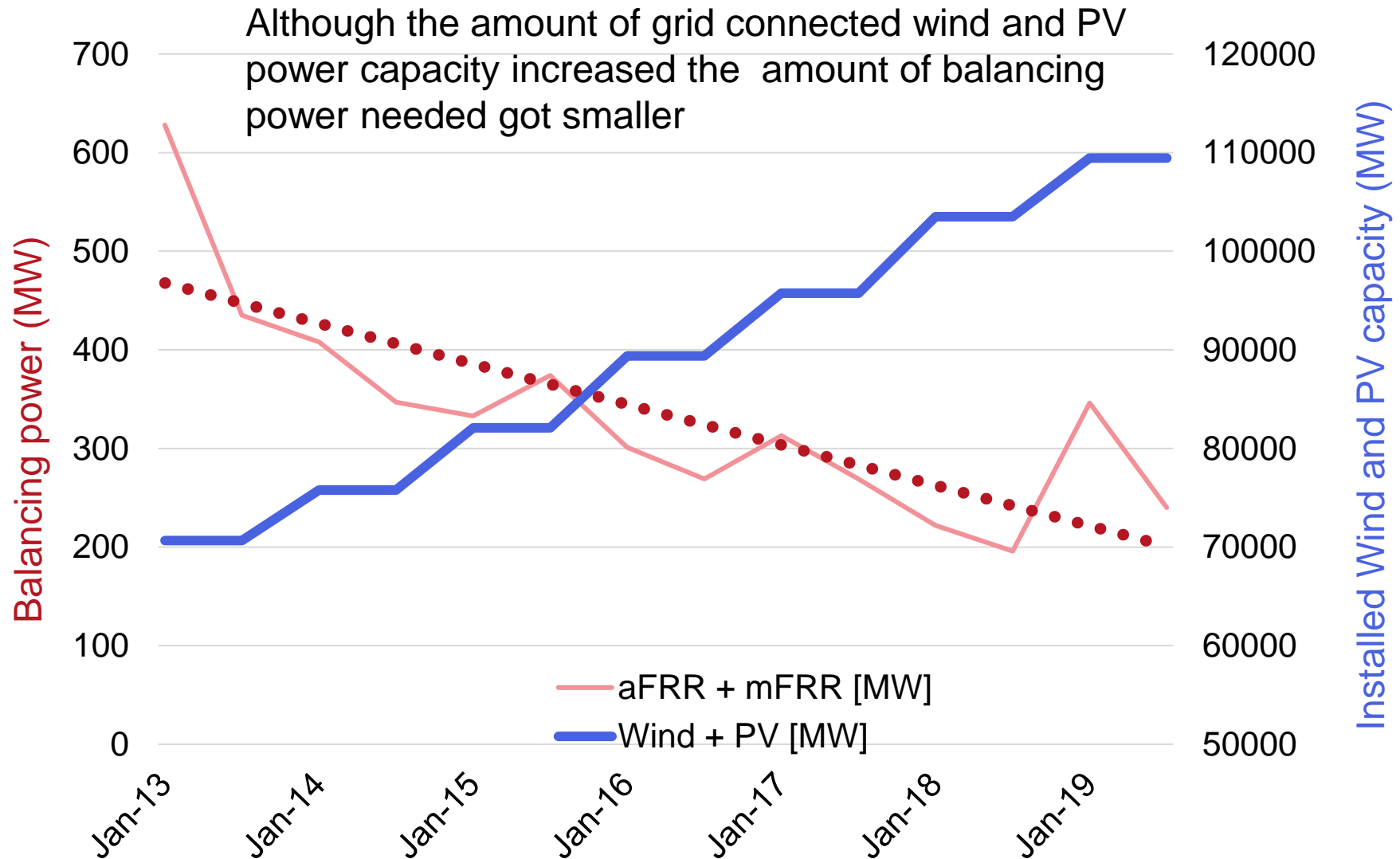


# Development of installed wind & PV capacity and used amount of balancing power (aFRR + mFRR) in Germany



Source: BNetzA „Monitoringbericht 2020“ and BMWi „Zeitreihen zur Entwicklung der erneuerbaren Energien in Deutschland“

# Development of installed wind & PV capacity and used amount of balancing power (aFRR + mFRR) in Germany



Source: BnetzA „Monitoringbericht 2020“ and BMWi „Zeitreihen zur Entwicklung der erneuerbaren Energien in Deutschland“

# Thank you!

**Albrecht Tiedemann**

**Renewables Academy (RENAC)**

Schönhauser Allee 10-11

D-10119 Berlin

Tel: +49 30 58 708 70 40

Fax: +49 30 58 708 70 88

tiedemann@renac.de



[www.renac.de](http://www.renac.de)